

A Stochastic Integral in Storage Theory

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1. Introduction

Throughout the following $\mathbf{R} = [0, +\infty)$, $\bar{\mathbf{R}} = [0, +\infty]$, $\mathbf{N} = \{0, 1, 2, \dots\}$, \mathcal{R} and $\bar{\mathcal{R}}$ the Borel subsets of \mathbf{R} and $\bar{\mathbf{R}}$ respectively, and \mathcal{R}^* the universally measurable subsets of $\bar{\mathbf{R}}$. If (E, \mathcal{E}) and (F, \mathcal{F}) are measurable spaces and $f: E \rightarrow F$ is measurable relative to \mathcal{E} and \mathcal{F} , then we write $f \in \mathcal{E}/\mathcal{F}$; if $(F, \mathcal{F}) = (\bar{\mathbf{R}}, \bar{\mathcal{R}})$ then we simply write $f \in \mathcal{E}$. By $\sigma(\cdot)$ we denote the σ -algebra generated by (\cdot) . If N is a transition probability from (E, \mathcal{E}) into (F, \mathcal{F}) , that is, $N(x, \cdot)$ is a probability on \mathcal{F} and $N(\cdot, A)$ is in \mathcal{E} , and $f \in \mathcal{F}$, then we write Nf for the function defined as

$$Nf(x) = \int N(x, dy) f(y), \quad x \in E.$$

Our notation and terminology follow that of [1].

Let (W, \mathcal{I}, P) be a probability space and let $\{A_t: t \geq 0\}$ be a process on W with stationary independent non-negative increments taking values in $(\mathbf{R}, \mathcal{R})$ with $P\{A_0 = 0\} = 1$. It is possible to, and we do, take $t \rightarrow A_t(w)$ to be right-continuous non-decreasing for each $w \in W$. Let $\mathcal{I}_t = \sigma(A_s: s \leq t)$. Throughout the first four sections we assume the jump rate to be finite.

Let $\Omega = \bar{\mathbf{R}} \times W$ and, for each probability measure μ on $\bar{\mathcal{R}}$, define P^μ to be the product measure $\mu \times P$ on $\bar{\mathcal{R}} \times \mathcal{I}$. In particular, when μ is the Dirac measure ε_x concentrated at $x \in \bar{\mathbf{R}}$, we write P^x for P^μ . Let $\mathbf{U} = \{P^\mu: \mu \text{ a probability on } \bar{\mathcal{R}}\}$, \mathcal{F} the completion of $\bar{\mathcal{R}} \times \mathcal{I}$ with respect to \mathbf{U} , and \mathcal{F}_t the completion of $\bar{\mathcal{R}} \times \mathcal{I}_t$ in \mathcal{F} with respect to \mathbf{U} .

For $\omega = (x, w) \in \Omega$, we shall write $A_t(\omega) = A_t(w)$ (there should be no confusion because of this) for all $t \geq 0$, and put $X_0(\omega) = x$.

Let r be a Lipschitz continuous strictly increasing function from $\bar{\mathbf{R}}$ into $\bar{\mathbf{R}}$ so that $r(0) = 0$, and consider the equation

$$(1.1) \quad X_t(\omega) = X_0(\omega) + A_t(\omega) - \int_0^t r(X_u(\omega)) du, \quad t \geq 0$$

for each $\omega \in \Omega$; (note that if $\omega = (+\infty, w)$, then $X_t(\omega) = +\infty$ for all $t \geq 0$). The reader will observe that certain of our results below go through under less stringent assumptions on r .

Ω is the sample space, \mathcal{F}_t the history until t . For the realization $\omega \in \Omega$, $X_0(\omega)$ is the initial content of the dam, $A_t(\omega)$ the total input during $[0, t]$, $X_t(\omega)$ the content at time t . The above equation states that the rate of output at time u is

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$r(X_u(\omega))$, and

$$(1.2) \quad Z_t(\omega) = \int_0^t r(X_u(\omega)) du$$

is the total output during $[0, t]$. $A = (\Omega, \mathcal{F}, \mathcal{F}_t, A_t, P^x)$ will be called the input process, $X = (\Omega, \mathcal{F}, \mathcal{F}_t, X_t, P^x)$ the content process, $Z = (\Omega, \mathcal{F}, \mathcal{F}_t, Z_t, P^x)$ the output process.

In the next section we shall show that the Eq. (1.1) has a unique solution for each ω and the resulting process $X = (\Omega, \mathcal{F}, \mathcal{F}_t, X_t, P^x)$ is a standard Markov process (a strong Markov process on a locally compact space with a countable base with $\mathcal{F}_t = \mathcal{F}_{t+} = \bar{\mathcal{F}}_t$ and which is right-continuous and quasi-left-continuous on $[0, +\infty)$).

The process $\{Z_t\}$ is a continuous additive functional of X and thus is completely characterized by its potential which we shall compute in Section 3. If we define

$$(1.3) \quad \tau_t(\omega) = \inf\{s: Z_s(\omega) > t\}$$

(then τ_t is the time the cumulative output reaches t) then $\tilde{X} = (\Omega, \mathcal{F}, \mathcal{F}_{\tau_t}, X_{\tau_t}, P^x)$ is also a standard Markov process. We shall compute its resolvent and thus specify its transition function. This also gives the distribution of $\tilde{A}_t = A_{\tau_t}$ since $X_{\tau_t} = X_0 + A_{\tau_t} - t$.

In Section 4 we consider the limiting distribution of the content process. We show that a limiting distribution for $\{X_t\}$ exists (and compute it) if

$$\sup_x r(x) > m$$

where m is the expected rate of input. If $\sup_x r(x) < m$, then $X_t \rightarrow +\infty$ as $t \rightarrow +\infty$.

In the special case where r is of form $r(x) = cx$ the equation can be solved explicitly in terms of A_t and, then, we are able to give a necessary and sufficient condition for the existence of the limiting distribution. This is put in Section 5.

The dam model considered was proposed by Moran [5]. We employ the methodology of [2] and [3] by considering the process X_t at its points of jumps along with the jump times (the resulting discrete time process is called a Markov renewal process). Section 3 is based on Chapters V and VI of Blumenthal and Gettoor [1]. Section 4 ties the results of Çinlar [2], Foguel [4], Orey [6], Pyke and Schaufele [7] to show the existence of a limiting distribution for the imbedded Markov renewal process from which the limiting distribution of X_t is computed directly.

2. Content Process

From the general theory of processes with stationary independent increments we can write

$$(2.1) \quad A_t(\omega) = at + \tilde{A}_t(\omega), \quad t \geq 0$$

where $a \geq 0$ is a constant and $\tilde{A}_t(\omega)$ is a right-continuous step function. We will denote the jump times by $\tau_1(\omega), \tau_2(\omega), \dots$, set $\tau_0(\omega) = 0$, and let $\alpha_1(\omega), \alpha_2(\omega), \dots$ be the magnitudes of the corresponding jumps.

It follows from (1.1) and (2.1) that $X_t(\omega)$ satisfies the differential equation

$$(2.2) \quad dx_t = a dt - r(x_t) dt$$

for $t \in (\tau_n(\omega), \tau_{n+1}(\omega))$ for any $n \in \mathbb{N}$. The following puts together all the relevant facts we need about (2.2). We omit the proof.

(2.3) **Lemma.** *The Eq. (2.2) with initial condition $x_0 = x$ has a unique solution $q(x, t)$.*

- a) For fixed $x \in \mathbb{R}$, $t \rightarrow q(x, t)$ is monotone continuous.
- b) For fixed $t \in \mathbb{R}$, $x \rightarrow q(x, t)$ is non-decreasing continuous.
- c) The mapping $(x, t) \rightarrow q(x, t)$ is in \mathcal{R}^2 .

$$(d) \quad -\frac{\partial}{\partial t} q(x, t) = [r(x) - a] \frac{\partial}{\partial x} q(x, t). \quad \square$$

(2.4) **Theorem.** *Define, for each $\omega \in \Omega$,*

$$\begin{aligned} \xi_0(\omega) &= X_0(\omega), \\ \xi_{n+1}(\omega) &= q(\xi_n(\omega), \tau_{n+1}(\omega) - \tau_n(\omega)) + \alpha_{n+1}(\omega), \quad n \in \mathbb{N} \end{aligned}$$

recursively. Then, we have

$$X_t(\omega) = q(\xi_n(\omega), t - \tau_n(\omega)) \quad \text{if } \tau_n(\omega) \leq t < \tau_{n+1}(\omega).$$

Proof. It follows from (2.1) that $X_t(\omega)$ satisfies (2.2) in any interval $(\tau_n(\omega), \tau_{n+1}(\omega))$, and at point $\tau_n(\omega)$, $X_t(\omega)$ jumps by $\alpha_n(\omega)$. Thus, if $\xi_n = X_{\tau_n}$, then X_t is as given above for $t \in [\tau_n, \tau_{n+1})$, and $\xi_{n+1} = X_{\tau_{n+1}-0} + \alpha_{n+1}$. \square

(2.5) **Theorem.** *$X = (\Omega, \mathcal{F}, \mathcal{F}_t, X_t, P^x)$ is a normal standard Markov process (and since the termination time is $+\infty$ a.s., trivially, a Hunt process).*

Proof. a) From the definition of X_0 and the measures P^x , $P^x\{X_0 = x\} = 1$ for any $x \in \mathbb{R}$. Thus we have normality trivially.

b) From the construction, $t \rightarrow X_t(\omega)$ is right-continuous and has left-hand limits by Lemma (2.3).

c) *Regularity Conditions.* Both X_t and A_t have the same jump times and the same magnitudes of jumps. Between the jumps, both are deterministic. Thus,

$$(2.6) \quad \mathcal{F}_t^0 \equiv \sigma(X_s: s \leq t) = \sigma(X_0, A_s: s \leq t) = \bar{\mathcal{R}} \times \mathcal{F}_t,$$

and therefore, \mathcal{F}_t is also the completion of \mathcal{F}_t^0 in \mathcal{F} relative to $\{P^\mu: \mu \text{ a probability on } \bar{\mathcal{R}}\}$. Thus,

$$(2.7) \quad X_t \in \mathcal{F}_t / \bar{\mathcal{R}}.$$

By right-continuity this implies that X is progressively measurable with respect to $\{\mathcal{F}_t\}$. Further, if T is any $\{\mathcal{F}_t\}$ stopping time,

$$(2.8) \quad X_T \in \mathcal{F}_T / \mathcal{R}^*,$$

since the \mathcal{F}_t are complete with respect to $\mathbf{U} = \{P^\mu: \mu \text{ a probability on } \bar{\mathcal{R}}\}$ and since the completion of \mathcal{F}_T in \mathcal{F} with respect to \mathbf{U} is the same as \mathcal{F}_T .

Finally, from the construction of P^x , for $A \in \bar{\mathcal{R}}$ and $B \in \mathcal{I}$, $P^x(A \times B) = \varepsilon_x(A)P(B)$ so that $x \rightarrow P^x(A \times B)$ is in $\bar{\mathcal{R}}$. By the monotone class theorem then

$$(2.9) \quad x \rightarrow P^x(A) \quad \text{is in } \bar{\mathcal{R}}$$

for any $A \in \bar{\mathcal{R}} \times \mathcal{I}$.

d) *Strong Markov Property.* Since $(W, \mathcal{F}, \mathcal{I}_t, A_t, P)$ is a process with stationary independent increments, $(W, \mathcal{F}, \mathcal{I}_{t+}, A_t, P)$ is a strong Markov process, and by (2.6), for any $\{\mathcal{F}_t^0\}$ stopping time T ,

$$(2.10) \quad P^x \{A_{T+t} - A_T \in B \mid \mathcal{F}_{T+}^0\} = P^x \{A_t \in B\}$$

for all $t \geq 0$, $B \in \mathcal{R}$ independent of x .

Let T be an $\{\mathcal{F}_t^0\}$ stopping time and define

$$(2.11) \quad A_t^+ = A_{T+t} - A_T, \quad X_t^+ = X_{T+t}, \quad t \geq 0.$$

Then, it follows from (1.1) that we have

$$(2.12) \quad X_t^+ = X_0^+ + A_t^+ - \int_0^t r(X_u^+) du, \quad t \geq 0$$

which is the same as (1.1). Since, by (2.10), $\sigma(A_t^+; t \geq 0)$ is independent of \mathcal{F}_{T+}^0 , this implies

$$P^x \{X_t^+ \in B \mid \mathcal{F}_{T+}^0\} = P^y \{X_t \in B\} \quad \text{on } \{X_0^+ = y\};$$

that is,

$$(2.13) \quad P^x \{X_{T+t} \in B \mid \mathcal{F}_{T+}^0\} = P^{X(T)} \{X_t \in B\}$$

for all $x \in \mathbf{R}$, $t \geq 0$, $B \in \mathcal{R}$.

This proves that $(\Omega, \mathcal{F}, \mathcal{F}_t^0, X_t, P^x)$ is a strong Markov process. By Proposition (8.12) in Chapter I of [1], then

$$(2.14) \quad \mathcal{F}_t = \mathcal{F}_{t+}, \quad t \geq 0.$$

Thus, by Theorem (7.3) in Chapter I of [1], if T is any $\{\mathcal{F}_t\}$ stopping time, for each μ there exists a $\{\mathcal{F}_{t+}^0\}$ stopping time T_μ such that $P^\mu \{T \neq T_\mu\} = 0$. Hence, (2.13) implies

$$(2.15) \quad P^x \{X_{T+t} \in B \mid \mathcal{F}_T\} = P^{X(T)} \{X_t \in B\}$$

for all $t \geq 0$, $x \in \mathbf{R}$, $B \in \mathcal{R}$, and $\{\mathcal{F}_t\}$ stopping times T .

e) If $\bar{\mathcal{F}}_t$ is the completion of \mathcal{F}_t in \mathcal{F} with respect to $\{P^\mu\}$, then obviously $\bar{\mathcal{F}}_t = \mathcal{F}_t$ and together with (2.14) we have

$$(2.16) \quad \mathcal{F}_t = \mathcal{F}_{t+} = \bar{\mathcal{F}}_t.$$

f) *Quasi-Left-Continuity.* Let $\{T_n\}$ be an increasing sequence of $\{\mathcal{F}_t\}$ stopping times with limit T . By the continuity of X_t on intervals (τ_n, τ_{n+1}) , $X_{T_n} \rightarrow X_T$ everywhere on Ω except on $A = \bigcup_{k=1}^\infty \{T = \tau_k\}$. Thus we need to show that $P^x(A) = 0$ for all x . Supposing otherwise, if $P^x(A) > 0$ for some x , then $P^x \{A_{T_n} \rightarrow A_T\} =$

$1 - P^x(A) < 1$ since the τ_k are discontinuity points for $\{A_t\}$ also. This contradicts the fact that a process with stationary independent increments is quasi-left-continuous. \square

We shall next derive the transition function

$$(2.17) \quad P_t(x, A) = P^x \{X_t \in A\}, \quad x \in \mathbf{R}, A \in \mathcal{R}$$

of the process X under the assumption that the rate of jumps is finite. Let, then, $b < \infty$ be the rate of jumps, and let γ be the distribution of the magnitude of a jump.

(2.18) **Lemma.** For any bounded $f \in \mathcal{R}^*$ and $t \geq 0$,

$$(2.19) \quad P_t f(x) = e^{-bt} f \circ q(x, t) + \int_0^t \int_0^\infty b e^{-bs} ds \gamma(dy) P_{t-s} f(y + q(x, s)).$$

(2.20) *Remark.* Since X is measurable, the mapping $(x, t) \rightarrow P_t(x, A)$ of \mathbf{R}^2 into $[0, 1]$ for fixed $A \in \mathcal{R}^*$ is $(\mathcal{R} \times \mathcal{R})^{\lambda \times \mu}$ -measurable for all finite measures λ and μ on \mathcal{R} (where the σ -algebra in question is the completion of $\mathcal{R} \times \mathcal{R}$ with respect to the product measure $\lambda \times \mu$). Thus, the integral on the right-hand side of (2.19) is well defined. \square

Proof. The first jump time $T = \tau_1$ is an $\{\mathcal{F}_t^0\}$, and therefore an $\{\mathcal{F}_t\}$, stopping time. Since X is strong Markov by Theorem (2.5),

$$P^x \{X_t \in B | \mathcal{F}_T\} = P_{t-T}(X_T, B) \quad \text{on } \{T \leq t\}.$$

By Theorem (2.4), $X_T = q(X_0, T) + \alpha_1$, and $X_t = q(X_0, t)$ on $\{T > t\}$. Thus, we have

$$P_t f(x) = E^x [f(X_t)] = E^x [I_{\{T > t\}} f \circ q(x, t) + I_{\{T \leq t\}} (P_{t-T} f)(\alpha_1 + q(x, T))]$$

which yields the lemma. \square

(2.21) **Theorem.** Let, for each $x \in \mathbf{R}, B \in \mathcal{R}^2$,

$$(2.22) \quad K(x, B) = \int \int b e^{-bs} ds \gamma(dy) I_B(q(x, s) + y, s)$$

and define

$$(2.23) \quad \begin{aligned} K_0(x, B) &= \varepsilon_{(x, 0)}(B) \\ K_{n+1}(x, B) &= \int K(x, d(y, s)) K_n(y, B - (0, s)). \end{aligned}$$

Then, for any bounded $f \in \mathcal{R}$,

$$(2.24) \quad P_t f(x) = \sum_{n=0}^\infty \int_{\mathbf{R} \times [0, t]} K_n(x, d(y, s)) e^{-b(t-s)} f \circ q(y, t-s)$$

exists and is the unique solution of (2.19).

Proof. Fix $f \in \mathcal{R}$ bounded and write $f(x, t) = P_t f(x)$ and $g(x, t) = e^{-bt} f \circ q(x, t)$. Then, (2.19) can be rewritten as

$$(2.25) \quad f(x, t) = g(x, t) + \int_{\mathbf{R} \times [0, t]} K(x, d(y, s)) f(y, t-s)$$

with K as defined by (2.22).

Note that $x \rightarrow K(x, B)$ is in \mathcal{R} and $B \rightarrow K(x, B)$ is a measure on \mathcal{R}^2 . Thus K is a semi-Markovian kernel and (2.25) is a Markov renewal equation (cf. Çinlar [2] and [3]). Then, that (2.24) exists and is a solution of (2.25) follows from Theorem 8 of [2] since

$$g(x, t) \leq K(x, \mathbf{R} \times (t, \infty)) = e^{-bt}.$$

That (2.24) is the only solution of (2.25) follows from Theorem (3.13) of [3] since

$$\sup_x K(x, \mathbf{R} \times [0, t]) = 1 - e^{-bt} < 1$$

for some $t > 0$ (because $b < \infty$ by hypothesis). \square

(2.26) *Remark.* Let $u(t, x) = e^{bt} P_t f(x)$ for some differentiable function f on (R, \mathcal{R}) . Then, from Lemma (2.18) we get

$$u(t, x) = f \circ q(x, t) + b \int_0^t \gamma(dy) \int_0^t u(s, y + q(x, t-s)) ds.$$

From this, using (d) of Lemma (2.3), we obtain

$$\left[\frac{\partial}{\partial t} + (r(x) - a) \frac{\partial}{\partial x} \right] u = b \int_0^\infty \gamma(dy) u(t, x + y).$$

This is the characteristic equation for X . \square

(2.27) *Remark.* Let K be as defined by (2.22) and put, for $\lambda \geq 0$,

$$K^\lambda(x, A) = \int_0^\infty e^{-\lambda t} K(x, A \times dt).$$

Then K^λ is a sub-Markovian kernel on (R, \mathcal{R}) . Putting

$$R^\lambda = \sum_n K_n^\lambda$$

we have

$$U^\lambda f(x) = \int_0^\infty e^{-\lambda t} P_t f(x) dt = \int_0^\infty R^\lambda(x, dy) \int_0^\infty e^{-(\lambda+b)t} f \circ q(y, t) dt$$

($U^\lambda f$ is called the λ -potential of f). This follows directly from Theorem (2.21) and the relationship between Laplace transforms and convolutions. \square

3. Output Process

Consider the process $\{Z_t; t \geq 0\}$ where

$$(3.1) \quad Z_t(\omega) = \int_0^t r(X_s(\omega)) ds, \quad \omega \in \Omega, t \geq 0.$$

Since X is right-continuous and r continuous, $t \rightarrow Z_t(\omega)$ is continuous. It is clear that, for any $s, t \geq 0$,

$$(3.2) \quad Z_{t+s} - Z_t \in \sigma(X_u: t \leq u \leq t+s),$$

and that $Z_0 = 0$. Thus, Z is a continuous additive functional of X .

The function

$$(3.3) \quad f_t(x) = E^x[Z_t]$$

is called the characteristic of Z . For fixed $x \in \mathbf{R}$, $f_0(x) = 0$, $t \rightarrow f_t(x)$ is non-decreasing continuous, and $f_t(x) = E^x[Z_t] \leq x + E^x[A_t] \leq x + mt$. Conditioning on \mathcal{F}_t , we have

$$(3.4) \quad f_{t+s}(x) = f_t(x) + \int P_t(x, dy) f_s(y),$$

or directly

$$(3.5) \quad f_t(x) = \int_0^t \int P_s(x, dy) r(y) ds.$$

The Laplace transform

$$(3.6) \quad u^\lambda(x) = \lambda \int_0^\infty e^{-\lambda t} f_t(x) dt, \quad \lambda \geq 0$$

is called the λ -potential of Z . From (3.5) and Remark (2.27) we have, in the notation of (2.27),

$$(3.7) \quad u^\lambda(x) = \int_0^\infty R^\lambda(x, dy) \int_0^\infty e^{-(\lambda+b)t} r \circ q(y, t).$$

We next consider the random time change effected by Z . Let

$$(3.8) \quad T_t(\omega) = \inf \{s: Z_s(\omega) > t\}, \quad t \geq 0$$

for each $\omega \in \Omega$. Since Z is continuous non-decreasing, T_t is strictly increasing right-continuous. For each $t \geq 0$, T_t is an $\{\mathcal{F}_s\}$ stopping time. Define

$$(3.9) \quad \bar{X}_t = X_{T_t}, \quad \bar{A}_t = A_{T_t}, \quad \bar{\mathcal{F}}_t = \mathcal{F}_{T_t}.$$

Noting that $Z_{T_t} = t$ by right-continuity of Z , (1.1) implies that

$$(3.10) \quad \bar{X}_t = X_0 + \bar{A}_t - t.$$

It is known (cf. [1] p. 212) that, since X is a standard Markov process and Z is continuous and strictly increasing (since r is Lipschitz, $X_t \neq 0$)

$$\bar{X} = (\Omega, \mathcal{F}, \bar{\mathcal{F}}_t, \bar{X}_t, P^x)$$

is also standard Markov process.

Further, for $x > 0$, $r(x) > 0$ and thus $P^x\{Z_t > 0\} = 1$ for $t > 0$ which implies that $P^x\{T_0 = 0\} = 1$ and hence $P^x\{\bar{X}_0 = x\} = 1$. On the other hand, if $x = 0$, $P^x\{T_0 = \tau_1\} = 1$ and thus $P^0\{\bar{X}_0 = \alpha_1 > 0\} = 1$. So, \bar{X} is normal except for $x = 0$.

Next we will compute the resolvent of \bar{X} (and thus specify its transition function, and by (3.10), that of \bar{A}_t).

(3.11) **Theorem.** *Let, for $\lambda > 0$, $f \in \mathcal{R}^*$ bounded,*

$$(3.12) \quad W^\lambda f(x) = E^x \left[\int_0^\infty e^{-\lambda t} f(\bar{X}_t) dt \right].$$

Then,

$$(3.13) \quad W^\lambda f(x) = e^{-\lambda x} \sum_{n=0}^\infty K_n g(x)$$

where

$$(3.14) \quad g(x) = \int_0^\infty \exp[-(b + \lambda a)t + \lambda q(x, t)] (f \cdot r) \circ q(x, t) dt$$

and

$$(3.15) \quad K(x, A) = \int_0^\infty b e^{-(b + \lambda a)t} dt \int_{A - q(x, t)} e^{-\lambda z} \gamma(dz),$$

for $x \in \mathbb{R}, A \in \mathcal{R}$.

Proof. Let $f \in \mathcal{B}^*$ be bounded, $\lambda > 0$. It follows from (3.8) and (3.9) that (cf. [1] Lemma (2.2) in Chapter V)

$$\int_0^\infty e^{-\lambda t} f(\bar{X}_t) dt = \int_0^\infty \exp(-\lambda Z_s) f(X_s) dZ_s.$$

Replacing Z_s by $X_0 + A_s - X_s$ and dZ_s by $r(X_s) ds$ and putting

$$(3.16) \quad h(x) = e^{-\lambda x} f(x) r(x),$$

we have

$$(3.17) \quad \begin{aligned} W^\lambda f(x) &= E^x \left[\exp(-\lambda X_0) \int_0^\infty \exp(-\lambda A_s) h(X_s) ds \right] \\ &= e^{-\lambda x} \int_0^\infty E^x [\exp(-\lambda A_s) h(X_s)] ds, \end{aligned}$$

where we use normality of X , and Fubini's theorem to change the order of integration.

Noting that $(\Omega, \mathcal{F}, \mathcal{F}_t, (X_t, A_t), P^x)$ has the strong Markov property, and that the first jump time $T = \tau_1$ is an $\{\mathcal{F}_t\}$ stopping time, we have

$$(3.18) \quad \begin{aligned} E^x [\exp(-\lambda A_t) h(X_t) | \mathcal{F}_T] \\ = \begin{cases} e^{-\lambda a t} h \circ q(x, t) & \text{on } \{T > t\}, \\ e^{-\lambda A(T)} E^{X(T)} [\exp(-\lambda A_{t-T}) h(X_{t-T})] & \text{on } \{T \leq t\}. \end{cases} \end{aligned}$$

Thus, if we write

$$(3.19) \quad k(x, t) = E^x [\exp(-\lambda A_t) h(X_t)], \quad k(x) = \int_0^\infty k(x, t) dt;$$

then we have from (3.18)

$$\begin{aligned} k(x, t) &= \exp(-b t - \lambda a t) h \circ q(x, t) \\ &\quad + \iint_{\mathbb{R} \times [0, t]} b e^{-b s} ds \gamma(dy - q(x, s)) \exp[-\lambda(a s + y - q(x, s))] k(y, t - s) \end{aligned}$$

which gives

$$\begin{aligned} k(x) &= \int_0^\infty \exp(-b t - \lambda a t) h \circ q(x, t) dt \\ &\quad + \int_0^\infty b e^{-b s} ds \int_0^\infty \gamma(dy - q(x, s)) \exp[-\lambda(a s + y - q(x, s))] k(y). \end{aligned}$$

Noting the definitions (3.14), (3.15), (3.16) this becomes

$$(3.20) \quad k = g + K k.$$

We have $K(x, \mathbf{R}) = \frac{b}{b + \lambda a} \gamma^x \leq \gamma^x < 1$ since $\lambda > 0$ where $\gamma^x = \int_0^\infty e^{-\lambda x} \gamma(dx)$. Therefore, $R = \sum_n K_n$ is well defined and $A \rightarrow R(x, A)$ is a finite measure. The solution of (3.20) then is

$$k = Rg.$$

This together with (3.19) and (3.17) yields (3.13). \square

4. Limit Theorems

In this section we shall consider the limiting distribution of X_t as t approaches infinity. First we want to show that, in the decomposition (2.1), we can take $a = 0$ without loss of generality.

Suppose $a > 0$ in (2.1). If $\sup_x r(x) < a$, then $t \rightarrow X_t(\omega)$ is a strictly increasing function and $X_t(\omega) \rightarrow +\infty$ as $t \rightarrow \infty$ for each $\omega \in \Omega$.

If $a > 0$ and $\sup_x r(x) = a$, then $t \rightarrow X_t(\omega)$ is non-decreasing, and there exists a hitting time T such that

$$X_t(\omega) = A_t(\omega) - at - C(\omega)$$

for all $t \geq T(\omega)$, (and $C(\omega)$ is independent of t). Therefore, then X_t and $A_t - at$ are shifted copies of each other; and $X_t(\omega) \rightarrow +\infty$ as $t \rightarrow \infty$ for each $\omega \in \Omega$.

Finally suppose $a > 0$, $\sup_x r(x) > a$, let x^* be such that $r(x^*) = a$ (since r is continuous, such a point exists), and set $T(\omega) = \inf\{t: X_t(\omega) > x^*\}$. Then, X_t is strictly increasing for $t < T$, and $X_{T+t} \geq x^*$, $r(X_{T+t}) \geq a$ for all $t \geq 0$. On the other hand, $P^x\{T < \infty\} = 1$ and $E^x(T) < \infty$ for all x . Thus, the set $[0, x^*)$ is distributive and $P^x\{X_t < x^*\} \rightarrow 0$ as $t \rightarrow \infty$ for $x < x^*$ (of course we have $P^x\{X_t < x^*\} = 0$ for $x \geq x^*$). Hence, it is sufficient to consider the limiting behavior of X restricted to $[x^*, \infty]$.

Then, putting $\tilde{r}(x) = r(x) - a$ we can rewrite (1.1) as $X_t = X_0 + \tilde{A}_t - \int_0^t \tilde{r}(X_u) du$.

Hence, from here on, we assume $a = 0$. We also assume the jump rate b to be finite, and denote by m the input rate:

$$m = b \int_0^\infty x \gamma(dx)$$

(then $E^x(A_t) = mt$ for all t).

Below we shall show that $P^x\{X_t \in B\} \rightarrow 0$ for all bounded B as $t \rightarrow \infty$ if $\sup_x r(x) < m$. If $\sup_x r(x) > m$, then $P^x\{X_t \in B\} \rightarrow \nu(B)$ as $t \rightarrow \infty$ for some probability measure ν independent of x and we will compute ν . The situation in the case $\sup_x r(x) = m$ is not clear. Our conjecture is that there exists a σ -finite (but not finite) invariant measure for P_t^x in that case.

Let η_n be the left-hand limit of X_t at τ_n , that is,

$$(4.1) \quad \begin{aligned} \eta_1 &= q(X_0, \tau_1) \\ \eta_{n+1} &= q(\eta_n + \alpha_n, \tau_{n+1} - \tau_n), \quad n = 1, 2, \dots, \end{aligned}$$

and put

$$(4.2) \quad \mathcal{M}_n = \sigma(X_0, \eta_1, \dots, \eta_n; \tau_1, \tau_2, \dots, \tau_n) \quad n = 1, 2, \dots$$

The following (whose proof we omit as fairly easy) is central to our development (cf. [2] for the definitions).

(4.3) **Proposition.** $(\Omega, \mathcal{F}, \mathcal{M}_n, (\eta_n, \tau_n), P^x)$ is a delayed Markov renewal process with semi-Markov kernel Q (that is,

$$P^x \{ \eta_{n+1} \in A, \tau_{n+1} - \tau_n \in B | \mathcal{M}_n \} = Q(\eta_n, A \times B)$$

for all $A, B \in \mathcal{B}$ and $n = 1, 2, \dots$), where

$$(4.4) \quad Q(x, \Gamma) = \iint b e^{-bs} ds \gamma(dy) I_\Gamma(q(x+y, s), s)$$

for all $x \in \mathbf{R}, \Gamma \in \mathcal{B}^2$.

(4.5) **Corollary.** $Y = (\Omega, \mathcal{F}, \mathcal{M}_n, \eta_n, P^x)$ is a Markov chain with transition kernel

$$(4.6) \quad N(x, A) = Q(x, A \times \mathbf{R}) = \iint b e^{-bs} ds \gamma(dy) I_A \circ q(x+y, s).$$

We next will show that the assumptions of Foguel [4] are satisfied by N and further that the σ -finite invariant measure ν he constructs for N is actually finite in our case.

Throughout the following we denote by T_f the operator defined as

$$T_f g(x) = f(x) g(x)$$

and write T_A instead of T_f when $f = I_A$. The following is of interest on its own.

(4.7) **Theorem.** Let $D = \{x: r(x) > c\}$ be non-empty for some constant $c > m$. Then,

- a) $(T_D N T_D)^n 1(x) \rightarrow 0$ as $n \rightarrow \infty$ for all $x \in \mathbf{R}$;
- b) $\sum_{n=0}^{\infty} (N T_D)^n 1(x) < \infty$ for all $x \in \mathbf{R}$;
- c) $\sum_{n=0}^{\infty} T_C (N T_D)^n 1$ is bounded for any bounded set $C \in \mathcal{B}$.

Proof. Note first that $(N T_D)^n = (N T_D)(T_D N T_D)^{n-1}$ so that b) implies a).

Since r is continuous non-decreasing, the set D is of form $D = (d, \infty)$. Thus, if $q(x, t) \in D$, then $q(x, t) < x - ct$ and $x - ct \in D$. From (4.1) therefore

$$(4.8) \quad \{ \eta_{k+1} \in D \} \subset \{ \eta_{k+1} \in D, \eta_{k+1} < \eta_k + \alpha_k - c(\tau_{k+1} - \tau_k) \in D \}.$$

Define

$$(4.9) \quad \begin{aligned} S_0 &= 0 \\ S_n &= \sum_{k=1}^n [\alpha_k - c(\tau_{k+1} - \tau_k)], \quad n = 1, 2, \dots \end{aligned}$$

Then, from (4.8) we have

$$(4.10) \quad \{ \eta_2 \in D, \dots, \eta_{n+1} \in D \} \subset \{ \eta_1 + S_1 \in D, \dots, \eta_1 + S_n \in D \} = \{ U_{\eta_1} > n \}$$

where

$$(4.11) \quad U_x = \inf \{ k \geq 1: x + S_k \notin D \}.$$

On the other hand, writing \tilde{P}^x for the conditional probability $P^y\{\cdot|\mathcal{M}_1\}$ on $\{\eta_1 = x\}$, we have

$$(4.12) \quad (NT_D)^n 1(x) = \tilde{P}^x\{\eta_2 \in D, \dots, \eta_{n+1} \in D\};$$

thus, from (4.10) and (4.11),

$$(4.13) \quad \sum_{n=0}^{\infty} (NT_D)^n 1(x) \leq \sum_{n=0}^{\infty} P^y\{U_x > n\} = E^y[U_x].$$

But $\{S_n\}$ is a random walk and the expected value of one of its steps is $m/b - c/b < 0$ since $c > m$. Thus, $\{S_n\}$ drifts to $-\infty$. From standard results for random walks, since D is of form (d, ∞) , $E^y[U_x] < \infty$ for any x . This proves b) via (4.13).

To show c) it is sufficient to take $C = [0, k]$. Then, the result is immediate from b) once we note that

$$\begin{aligned} \sup_x \sum_{n=0}^{\infty} T_C (NT_D)^n 1(x) &= \sup_{x \in C} \sum_{n=0}^{\infty} (NT_D)^n 1(x) \\ &\leq \sup_{x \in C} E^y(U_x) \leq E^y(U_k). \quad \square \end{aligned}$$

(4.14) **Lemma.** *If $f \in \mathcal{R}$ is continuous, so is Nf .*

Proof. From Corollary (4.5)

$$Nf(x) = \iint b e^{-bs} ds \gamma(dy) f \circ q(x + y, s)$$

and this is continuous since $x \rightarrow q(x, t)$ is continuous by Lemma (2.3). \square

(4.15) **Theorem.** *If $\sup_x r(x) > m$, then Y has a unique invariant measure ν and $\nu(\mathbf{R}) = 1$.*

Proof. By the hypothesis of the theorem, there exists $c > m$ with $D = \{x: r(x) > c\}$ non-empty. Thus, Theorem (4.7) holds. The statement a) of (4.7) is assumption (2.1) of Foguel [4], and Lemma (4.14) is assumption (3.5) in [4]. Thus all the results of [4] hold for the process Y . In particular, the theorem of that paper shows that there exists a measure ν for Y which is constructed as follows.

Let C be a closed interval containing the complement $\mathbf{C} D$ of D ; and let f be a continuous function with

$$0 \leq f \leq 1, \quad f = 0 \quad \text{on } \mathbf{C} D, \quad f = 1 \quad \text{on } \mathbf{C} C.$$

Then

$$N_\infty = \sum_{n=0}^{\infty} (NT_f)^n NT_{1-f}$$

is a well defined contraction on bounded measurable functions defined on C . Further, by the corollary to Lemma 3 of [4], there exists a probability measure λ , on C , with $\lambda N_\infty = \lambda$. We have

$$(4.16) \quad \nu = \sum_{n=0}^{\infty} \lambda (NT_f)^n$$

as the desired σ -finite measure satisfying $\nu N = \nu$.

Since $f \leq I_D$, $NT_f \leq NT_D$ and by iteration we have $(NT_f)^n \leq (NT_D)^n$. Thus, from (4.16),

$$v 1 \leq \lambda \sum_{n=0}^{\infty} (NT_D)^n 1 \leq \sup_{x \in C} \sum_{n=0}^{\infty} (NT_D)^n 1(x)$$

since λ is a probability measure concentrated on C . Hence $v 1 < \infty$ by c) of Theorem (4.7). Then we can take $v(\mathbf{R}) = 1$ by a suitable normalization.

To show that v is the only invariant measure we note that, for any $x \in \mathbf{R}$ and $A \in \mathcal{R}$ with positive Lebesgue measure, there exists n such that $N_n(x, A) > 0$. This follows from the nature of the exponential distribution since, if $\beta = \sup \{x: \gamma(0, x) < 1\}$ and $A = [a_0, a_1] \subset [0, x + \beta - \varepsilon]$, then

$$N(x, A) \geq \int_{\beta - \varepsilon}^{\beta} \gamma(dy) \int_{B_y} b e^{-bs} ds > 0$$

where $B_y = \{t: q(x + y, t) \in A\}$. \square

(4.17) **Theorem.** Let $N(t) = \sup \{n: \tau_n \leq t\}$. Then, if $\sup_x r(x) > m$,

$$(4.18) \quad \lambda(A \times B) = \lim_{t \rightarrow \infty} P^x \{ \eta_{N(t)} \in A, t - \tau_{N(t)} \in B \}$$

exists for open sets A, B and is given by

$$(4.19) \quad \lambda(A \times B) = v(A) \int_B b e^{-bs} ds$$

where v is as defined in Theorem (4.15).

Proof. Existence of the limit (4.18) is assured by Theorem 4 of Orey [6] whose conditions we satisfy as follows. Conditions (i) and (ii) of [6] on the state space are satisfied by $(\mathbf{R}, \mathcal{R})$; condition (iii) is satisfied by our Theorem (4.15); and the finiteness of the integral (0.1) of [6] is evident since v is finite and the expected sojourn times involved are $1/b$.

That (4.19) is true follows from Theorem (3.1) of Pyke and Schaufele [7] (though their state space was discrete, their proof goes through for our case).

(4.20) **Theorem.** If $\sup r(x) > m$, then

$$\mu(A) = \lim_{t \rightarrow \infty} P^x \{ X_t \in A \}$$

exists and we have

$$\mu(A) = \iiint v(dx) \gamma(dy) b e^{-bs} ds I_A \circ q(x + y, s).$$

Proof. From Theorem (2.4) and (4.1) we have

$$X_t = q(\eta_n + \alpha_n, t - \tau_n) \quad \text{on } \{ \tau_n \leq t < \tau_{n+1} \},$$

that is, if $N(t)$ is defined as in Theorem (4.17),

$$X_t = q(\eta_{N(t)} + \alpha_{N(t)}, t - \tau_{N(t)}).$$

Proof follows now from Theorem (4.17) and the independence of $\eta_n, \alpha_n, \tau_{n+1} - \tau_n$ for each n . \square

(4.21) *Remark.* Suppose $\sup_x r(x) = c < m$. Then, from (1.1) we have

$$X_t \geq X_0 + A_t - ct.$$

Since $E^x[A_t] = mt > ct$, this implies that $\lim_t X_t = +\infty$ almost surely for any P^x .

5. A Special Case

Consider the case $r(x) = cx$ where $c > 0$ is a constant. According to Theorem (4.20), a limiting distribution for X exists if m is finite. We will now show by an explicit calculation that it may still exist when m is infinite. Furthermore, in this case we do not need to assume the rate of jumps to be finite. For the reasons explained in Section 4 we assume, without loss of generality, that $\{A_t\}$ is a pure jump process. Then,

$$(5.1) \quad E^x[\exp(-\lambda A_t)] = \exp(-tg(\lambda))$$

where

$$(5.2) \quad g(\lambda) = \int_0^\infty (1 - e^{-\lambda x}) \nu(dx)$$

for some measure ν on the Borel subsets of $(0, \infty)$ satisfying $\int (x/(1+x)) \nu(dx) < \infty$.

(5.3) **Theorem.** *Let $r(x) = cx$. In order that X_t have a limiting distribution when t approaches $+\infty$, it is necessary and sufficient that*

$$(5.4) \quad \int_1^\infty (\log x) \nu(dx) < \infty.$$

Proof. In this case the Eq. (1.1) can be solved explicitly to yield

$$(5.5) \quad X_t(\omega) = X_0(\omega) e^{-ct} + \int_0^t e^{-c(t-s)} dA_s(\omega)$$

where the integral on the right-hand side is a Riemann-Stieltjes integral. If we write (5.5) as a limit of Riemann-Stieltjes sums, it follows from (5.1) that

$$\begin{aligned} u_t(x, \lambda) &= E^x[e^{-\lambda X_t}] \\ &= \exp(-\lambda x e^{-ct}) \exp\left[-\int_0^t g(\lambda e^{-cs}) ds\right]. \end{aligned}$$

If we now make the change of variable $y = \lambda e^{-cs}$ above, it becomes clear that $\lim_{t \rightarrow \infty} u_t(x, \lambda)$ exists simultaneously with

$$(5.6) \quad \lim_{\epsilon \downarrow 0} \int_\epsilon^\lambda \frac{1}{y} g(y) dy.$$

But this last expression is of the form $\int_0^\infty f_\epsilon(x) \nu(dx)$ where

$$f_\epsilon(x) = \int_\epsilon^\lambda \frac{1 - e^{-yx}}{y} dy.$$

As $\varepsilon \downarrow 0$, $f_\varepsilon(x)$ increases to the limit

$$(5.7) \quad f(x) = \int_0^\lambda \frac{1 - e^{-yx}}{y} dy.$$

Thus, by the monotone convergence theorem, the limit (5.6) exists simultaneously with the integral

$$(5.8) \quad \int_0^\infty f(x) \nu(dx).$$

But f is locally bounded and asymptotic to $\log x$ as $x \rightarrow \infty$. Therefore, the conclusion follows from the continuity theorem for Laplace transforms. \square

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