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#### On the reconstruction of a killed Markov process

#### By Sadao Sato

#### 1. Introduction

Let  $(Y_t)$  be a right continuous strong Markov process and  $(S_t)$  be its semigroup. Let  $(M_t)$  be a multiplicative functional satisfying  $\not\subseteq M_t \subseteq I$ . Then the formula

(1.1) 
$$T_{t}f(x)=E_{x}[f(Y_{t})M_{t}]$$

defines a subordinated semigroup. P. A. Meyer obtained the existence and the uniqueness (up to the life time of  $Y_t$ ) of such an  $M_t$  for every positive semigroup subordinated to  $(S_t)$  (see [DM3]). A Markov process with semigroup  $(T_t)$  is called (according to Dynkin) a subprocess of  $(Y_t)$ . It is a strong Markov process if the functional  $(M_t)$  is exact, and in particular if  $M_0=1$  a.s..

Conversely, let  $(X_t)$  be a right continuous strong Markov process with semigroup  $(T_t)$ . The problem in this paper is to construct in a natural way a strong Markov process  $(Y_t)$  with semigroup  $(S_t)$  such that 1)  $(X_t)$  is subordinated to  $(Y_t)$  2)  $(Y_t)$  is as close as possible to being conservative. This problem was studied by several authors. In Ikeda, Nagasawa and Watanabe [INW], it is constructed using the "piecing out method". They also proved that it is conservative under the condition

(U) 
$$\inf_{\mathbf{x} \in S} \{ P_{\mathbf{x}}[\xi > t] \} > 0 \text{ for some t>0,}$$

which means the "uniformity of the killing". P. A. Meyer[M2] obtained that the piecing out method preserves the class of right processes.

The natural idea to reconstruct  $\boldsymbol{S}_t$  is to try to give a meaning to the formula

(1.2) 
$$S_t f(x) = E_x^T [f(X_t) \frac{1}{M_+}].$$

However, the expectation being relative to the small semigroup  $(T_t)$ , we must find a way to describe the MF  $(M_t)$  on the sample space of  $(X_t)$ . The precise description in the general case will be given later, but there is a simple particular case, when X has a totally inaccessible lifetime 5. Then the decreasing process  $S_t = I_{\{t < S\}}$  has a continuous predictable compensator  $A_t$  which is a continuous additive functional of X. Then the appropriate version of the MF  $1/M_t$  in (1.2) is  $\exp(A_t)$ . In the general case, we must use "Stieltjes exponentials" in the sense of Sharpe[S] instead of ordinary exponentials. We can prove in all cases that the formula gives a larger semigroup  $(S_t)$  and still submarkov. However, it is not necessarily conservative. We can prove it is conservative if we have "uniformity of the killing" and the totally inaccessibility of S (Theorem 4.5).

In Section 2, we justify the formula (1.2) in the relative theory of  $(Y_t)$  and  $(X_t)$ . We give an example which is important in Section 4 and 5.

Next, we forget  $(Y_t)$  and then we can see only  $(X_t)$ . In Section 3, we obtain a general formula of 1/M. We can get a theorem concerning the continuity of a compensator.

In Section 4, we study the conservativity of  $(S_t)$ .

In Section 5, we only assume the totally inaccessibility of the life time. Now,  $(S_t)$  is not conservative in general. However, we can still get a conservative semigroup in a weak sense. We also show some properties of  $(S_t)$ .

#### 2. Basic results on the killing

In this section, we recall the basic construction of the killing by a decreasing right continuous multiplicative functional and point out some facts which suggest our main results in latter sections.

The construction of a subprocess is discussed by many authors. We describe it along the method in Blumenthal-Getoor[BG].

Let  $(Y_t, 2, 3, 3, t, P_x, 5)$  be a right continuous strong Markov process and  $(S_t)$  be its semigroup. We denote the state space by E, which is locally compact metrizable, and the death point by  $\partial$ . For every function f on E, we always set  $f(\partial)=0$  for convenience.

Let  $(M_t)$  be a right continuous MF satisfying  $0 \le M_t \le 1$ . Let  $E_M$  be the set of permanent points, that is  $E_M = \{x \in E; P_X[M_0 = 1] = 1\}$  which is universally measurable. If x is not permanent, then  $P_X[M_0 = 0] = 1$ .

We consider the product space  $\overline{\mathcal{Q}} = [0, \omega] \times \mathcal{Q}$  and denote its element by  $(r, \omega)$ . We denote the coordinate map from  $\overline{\mathcal{Q}}$  to  $[0, \omega]$  by  $R(r, \omega) = r$ . Define

(2.1) 
$$X_{+}(r,\omega)=Y_{+}(\omega)$$
 if t\partial if not.

Thus the life time of  $(X_t)$  is  $\overline{\zeta} = \zeta \wedge R$ . Define the translation operator by

(2.2) 
$$\overline{\theta}_{t}(r,\omega) = ((r-t)^{+}, \theta_{t}\omega),$$

which guarantees  $\overline{\theta}_t \circ \overline{\theta}_s = \overline{\theta}_{t+s}$  and  $X_s \circ \overline{\theta}_t = X_{t+s}$ . We give the filtration  $\overline{\mathcal{F}}_t$  generated by  $\mathcal{F}_t$ -measurable random variables (considered as a variable on  $\overline{\Omega}$  not depending on r) and  $R \wedge t$ .

Define

Then this filtration is right continuous and  $\mathcal{F}_t\supset \overline{\mathcal{F}}_t$ . Note that if g is  $\mathcal{F}_t$ -measurable, then there exists a  $\mathcal{F}_t$ -measurable h such that g=h on  $\{t < R\}$ .

Define the probability measure  $\overline{P}_Y$  on  $\overline{Q}$  by

(2.4) 
$$\overline{E}_{x}[f(r,\omega)] = E_{x}[-\int_{[0,\infty]} f(r,\omega) dM_{r}(\omega)],$$

where let  $M_{0-}=1$ ,  $M_{\infty}=0$ . If x is not permanent, we have  $\overline{P}_{x}(R=0)=1$ .

For every  $g_t$ -measurable  $f_t$ , we have by (2.4)

$$(2.5) \qquad \overline{\mathbb{E}}_{\mathbf{x}}[\mathbf{f}_{\mathsf{t}} \mathbf{1}_{\{\mathsf{t} < \mathsf{R}\}}] = \mathbb{E}_{\mathbf{x}}[-\mathbf{f}_{\mathsf{t}}(\omega)] (\mathbf{t}, \omega) dM_{\mathbf{r}}(\omega) = \mathbb{E}_{\mathbf{x}}[\mathbf{f}_{\mathsf{t}} \mathbf{M}_{\mathsf{t}}].$$

Specially, for every measurable function f (remembering that  $f(\partial)=0$ ), we have

(2.6) 
$$T_{t}f(x)=\overline{E}_{x}[f(X_{t})]=E_{x}[f(Y_{t})M_{t}],$$

where  $T_t$  denotes the semigroup of  $(X_t)$ . We can easily obtain the Markov property of  $(X_t)$ :

$$\overline{\mathbb{E}}_{\mathbf{X}}[\mathbf{f}(\mathbf{X}_{\mathsf{t}})|\boldsymbol{\mathcal{J}}_{\mathsf{S}}] = \mathbf{T}_{\mathsf{t-S}}\mathbf{f}(\mathbf{X}_{\mathsf{S}}).$$

Remark 2.1. We can construct  $(X_t)$  by considering the product of  $(Y_t)$  and an independent variable  $\epsilon$  which has the exponential distribution with parameter one and define

$$R=\inf\{t;-\log(M_t)>e\}.$$

This construction (due to Hunt) is more intuitive than the above. However, it becomes difficult to define the translation operator and get the Markov property. See Azema[A] for the "relative theory" of a general process.

The value of  $M_t$  in  $(\zeta, \omega)$  has no meaning for the subprocess. Thus we can take a normalization of  $(M_t)$  by setting

(2.7) 
$$M_t=0$$
 for  $t \ge \zeta$ ,

which we will assume in the following. Another selection is to set  $M_t = M_{\zeta}$  for  $\geq \zeta$ . Define

(2.8) 
$$\rho = \inf\{t; M_t = 0\},$$

which is a terminal time for  $(X_t)$ . We have  $\rho \leq \zeta$  under the condition (2.7).

Theorem 2.2. We can invert (2.6) as

(2.9) 
$$\mathbb{E}_{\mathbf{X}}[f(Y_{\mathsf{t}});\mathsf{t} < \rho] = \overline{\mathbb{E}}_{\mathbf{X}}[f(X_{\mathsf{t}}) \frac{1}{M_{\mathsf{t}}}].$$

Thus we can reconstruct the larger semigroup from  $X_t$  by  $1/M_t$  iff  $P \ge 5$ .

Proof. By (2.4), we obtain R < P on  $\{R < \omega\}$   $\overline{P}_X$ -a.s.. Thus  $M_t > 0$  a.s. for t < R. By (2.5), we get

$$\overline{E}_{x}[1_{\{t \leq R\}}f(X_{t})/M_{t}] = E_{x}[f(Y_{t})/M_{t} \cdot M_{t};M_{t}>0] = E_{x}[f(Y_{t});t<\rho].$$

Remark 2.3. (a) If  $M_t$  is equal to zero with a positive probability on  $\{t < S\}$ , we can not reconstruct the upper process  $(Y_t)$  completely. Typically,  $M_t$  becomes zero at an absorbing boundary. Then we cannot avoid to change the state space or to give an external condition (like an entrance law) to get a conservative upper process in general.

(b) Under the normalization (2.7), since  $R \lt P \le 5$  on  $\{R \lt \omega\}$ , we have  $R = \overline{5}$  on  $\{R \lt \omega\}$   $\overline{P}_X$ -a.s..

Now, we introduce the increasing process  $(A_t)$  by the "Stieltjes logarithm":

(2.10) 
$$-dM_{s}/M_{s}-dA_{s}$$
 on  $[0,\omega]$ ,

with the condition  $A_{0}$ =0. Define

(2.11) 
$$a(t,\omega) = -\log(M_{+}(\omega)),$$

which is a  $[0,\infty]$ -valued AF. Let  $(a_t^C)$  be the continuous part of  $(a_t)$  and denote the jump by  $\Delta a_S = a_S - a_S$ . Then we can write

(2.12) 
$$A_{t} = \begin{cases} a_{t}^{C} + \sum_{s \leq t} \{1 - \exp(-\Delta a_{s})\} \text{ on } [0, \rho), \\ A_{\rho} + 1_{\{M_{\rho} > 0\}} \text{ on } [\rho, \omega]. \end{cases}$$

Note that  $\Delta A_t < 1$  when  $t < \rho$  and  $\Delta A_{\rho} = 1$  if  $M_{\rho} > 0$  ( $\rho$  may be  $\omega$ ). For details, see Sharpe[S] and Meyer[M1].

We consider  $A_t$  as a random variable on  $\overline{\mathcal{Q}}$ . Since  $A_t(r,\omega)=A_t(\omega)$  by our rule, we have  $A_{t\Lambda\,R}(r,\omega)=A_{t\Lambda\,r}(\omega)$ , which is  $\mathcal{F}_t$ -measurable. We also note that

 $A_R(r,\omega)=A_r(\omega)$  is  ${\mathcal F}$ -measurable. Define a right continuous process on  ${\overline {\mathcal Q}}$  by

$$(2.13) m_t = 1 \{ t < \overline{\xi} \}^{+A} t_{\Lambda} \overline{\xi}.$$

Since  $\overline{\zeta} \subseteq R \not\hookrightarrow$  on  $\{R \not\hookrightarrow_{\chi}\}$   $\overline{P}_{\chi}$ -a.s.,  $(A_{t,\chi}\overline{\zeta})$  is a finite right continuous AF.

Lemma 2.4.

$$(2.14) \overline{E}_{x}[A_{R}]=1.$$

(2.15) 
$$\overline{E}_{\mathbf{x}}[\mathbf{m}_{\mathsf{t}}]=1.$$

*Proof.* Since  $A_t$  and  $M_t$  are of finite variation, by (2.10) we obtain  $d(A_rM_r)=M_{r-}dA_r+A_rdM_r=-dM_r+A_rdM_r \text{ in } [0,\infty]. \text{ Therefore, by (2.4) we obtain}$ 

$$\overline{E}_{x}[A_{R}] = E_{x}[-\int_{[0,\infty]} A_{r}(\omega) dM_{r}(\omega)] = E_{x}[-\int_{[0,\infty]} d(A_{r}M_{r} + M_{r})]$$

$$= E_{x}[-A_{m}M_{m} + A_{0} - M_{0} - M_{m} + M_{0}] = 1.$$

Similarly, we get

(2.16) 
$$\overline{E}_{\mathbf{x}}[A_{\mathsf{t},\mathsf{A}}R] = \mathbf{E}_{\mathbf{x}}[-\int_{[0,t]}A_{\mathsf{r}}dM_{\mathsf{r}} - \int_{(t,\omega)}A_{\mathsf{t}}dM_{\mathsf{r}}]$$
$$= \mathbf{E}_{\mathbf{x}}[-A_{\mathsf{t}}M_{\mathsf{t}} - M_{\mathsf{t}} + M_{\mathsf{O}_{\mathsf{c}}}] - \mathbf{E}_{\mathbf{x}}[A_{\mathsf{t}}(M_{\mathsf{m}} - M_{\mathsf{t}})] = 1 - \mathbf{E}_{\mathbf{x}}[M_{\mathsf{t}}].$$

On the other hand, we have

$$(2.17) \overline{E}_{\mathbf{x}}[1_{\{t<\overline{\zeta}\}}] = T_{t}1(\mathbf{x}) = E_{\mathbf{x}}[M_{t};t<\zeta] = E_{\mathbf{x}}[M_{t}],$$

where we used (2.7). By the definition of  $A_t$ , we have  $A_{t\Lambda R} = A_{t\Lambda \overline{\xi}}$ . Adding (2.16) and (2.17), we conclude (2.15).

For the simplicity, we write  $\overline{\xi}_{t}=1_{\{t<\overline{\xi}\}}$  and  $\overline{A}_{t}=A_{t}$  Since  $(\overline{A}_{s})$  is an AF, we get

$$\overline{\mathbb{E}}_{\mathbf{x}}[\overline{\mathbb{F}}_{\mathsf{t+s}} + \overline{\mathbb{A}}_{\mathsf{t+s}} | \mathcal{J}_{\mathbf{s}}] = \overline{\mathbb{A}}_{\mathbf{s}} + \overline{\mathbb{E}}_{\overline{\mathbb{X}}_{\mathbf{s}}}[\overline{\mathbb{F}}_{\mathsf{t}} + \overline{\mathbb{A}}_{\mathsf{t}}] = \overline{\mathbb{A}}_{\mathbf{s}} + \overline{\mathbb{F}}_{\mathbf{s}} = m_{\mathbf{s}}.$$

Therefore we obtain

Theorem 2.5. The process  $(m_t)$  defined by (2.13) is a martingale with respect to  $(\overline{P}_x \mathcal{J}_t)$ .

Corollary 2.6. Assume that  $(M_t)$  is continuous in [0,P) and one of the following conditions is satisfied:

- (a)  $M_t > 0$  on  $\{t < \xi\}$  and  $M_{\xi} = 0$  a.s.
- (b)  $(Y_t)$  is conservative and  $M_t > 0$  a.s..

Then  $\overline{A}_t$  is a PCAF(positive continuous AF) and the life time  $\overline{\xi}$  of  $(X_t)$  is totally inaccessible. Moreover, we have

(2.18) 
$$S_{t}f(x)=\overline{E}_{x}[f(X_{t})\exp(\overline{A}_{t})].$$

*Proof.* Under the condition (a) or (b), we see that  $\rho=5$  and  $\overline{A}_t$  has no jumps by (2.12).

Proposition 2.7. Assume that  $(M_t)$  is continuous in [0,P) and  $M_{P-}=0$  almost surely. Then the variable  $\overline{A}_{\overline{\zeta}}$  has the exponential distribution with parameter one.

*Proof.* Fix any  $\geq 0$  and define  $T=\inf\{t>0; a_t>c\}$ . By the continuity of  $a_t$ , we have

$$\overline{P}_{\mathbf{x}}[\overline{A}_{\overline{\zeta}} > c] = \overline{P}_{\mathbf{x}}[R > T] = E_{\mathbf{x}}[-\int_{[T, \infty]} dM_{\mathbf{r}}(\omega)] = E_{\mathbf{x}}[M_{T}] = e^{-C}.$$

Now, we give an example which suggests our situation to be treated in Section 4 and 5.

**Example.** Let  $(Y_t)$  be the standard Brownian motion on  $(0, \infty)$  with the absorbing boundary at zero. We define

(2.19) 
$$a_t = \int_0^t Y_s^{-2} ds$$
,

and  $(X_t)$  be the subordinated process by  $M_t = \exp(-a_t)$ . Since  $a_{\xi} = \infty$  almost surely, by Corollary 2.6(a), we have

$$(2.20) \overline{A}_t = \int_0^{t \wedge \zeta} X_s^{-2} ds$$

and  $\overline{\xi}$  is totally inaccessible. By the inverse formula (2.18), we can get

 $(Y_t)$  from  $(X_t)$  by  $exp(\overline{A}_t)$ . But,  $(Y_t)$  is not conservative. In addition, we emphasize that the life time of  $(Y_t)$  is predictable and the Doob-Meyer decomposition of  $S_t$  like (2.13) is trivial.

#### 3. Inversion of the killing

In this section, we fix a right continuous strong Markov process  $(X_t, \Omega, \mathcal{F}, P_x, S)$ . Our aim is to extend its semigroup  $(T_t)$ .

Let  $(N_t)$  be a right continuous MF satisfying  $N_t \ge 1$  (necessarily  $N_0 = 1$ ). Define a semigroup  $(S_t)$  by

(3.1) 
$$S_{+}f(x)=E_{x}[N_{+}f(X_{+})].$$

Clearly, we have  $S_t \ge T_t$ . Let  $S_t = I_{\{t \le S_t\}}$ . It is easy to see that the process  $(N_t S_t)$  is a supermartingale iff  $(S_t)$  is submarkov.

Let  $(A_t)$  be any AF such that the process

(3.2) 
$$m_{+} = \zeta_{+} + A_{+}$$

is a supermartingale. Since  $(N_t)$  and  $(\xi_t)$  are of finite variation, we have

(3.3) 
$$d(N_t \xi_t) = N_{t-} d\xi_t + \xi_t dN_t = N_{t-} dm_t - N_{t-} dA_t + \xi_t dN_t.$$

We introduce the "Stieltjes exponential" of  $(A_t)$ , that is the unique solution of

(3.4) 
$$d\hat{N}_{+}/\hat{N}_{+}=dA_{+} \text{ on } [0,\infty) \text{ with } \hat{N}_{0}=1,$$

which is a MF and its value is

(3.5) 
$$N_t = \exp(A_t^C) \prod_{s \le t} (1 + \Delta A_s).$$

Since the supermartingale  $(m_t)$  has finite variation and its jumps are  $\geq -1$ , the Doléans exponential  $D_t$  of  $m_t - m_0$  is a positive supermartingale and its expectaion is  $\leq 1$ . By the Doléans formula, we have

$$(3.6) D_t = N_t if t < \zeta, = D_{\zeta} - \Delta A_{\zeta} if t \ge \zeta,$$

which is a MF. Putting  $M_t = D_t \xi_t$ , we get a normalized MF which still has

expectation  $\leq 1$ . Hence we obtain

Theorem 3.1. The semigroup  $(S_t)$  associated with  $(N_t)$  is submarkov and an extension of  $(T_t)$ .

Corollary 3.2. Let  $(A_t^C)$  be the continuous part of  $(A_t)$ . Then

(3.7) 
$$S_{t}f(x)=E_{x}[f(X_{t})exp(A_{t}^{C})]$$

is an extended submarkov semigroup.

Remark 3.3. If  $T < \xi$  is a stopping time, we have  $E_X[A_T - A_O] \leq E_X[\xi_O - \xi_T] = 0$ , and therefore  $A_T = 0$ . Thus if  $\xi$  is predictable, this extension is completely trivial.

Now we consider a (unique) predictable AF  $(A_t)$  which generates the pure excessive part  $j_0$  of the function j=1 on E. Since  $j(X_t)=\xi_t$ , (3.2) gives the Doob-Meyer decomposition and  $(m_t)$  is a martingale of the class  $X^t$ . Since  $(N_{t-})$  is predictable, the first term of the right side of (3.3) is a local martingale.

In the following, we consider the case that  $(A_t)$  is predictable and  $(m_t)$  is a martingale. We do all things in the general theory of stochastic processes. First, we recall

Lemma 3.4 Let  $(A_t)$  be a predictable increasing process.

(a) For every nonnegative function  $\phi$  on  $[0,\infty)$  with  $\phi(0)=0$ , define

(3.8) 
$$C_{t} = A_{t}^{C} + \sum_{s \leq t} \phi (\Delta A_{s}).$$

Then  $(C_t)$  is predictable.

(b) If  $(A_t)$  is continuous on  $(T, \omega)$  for a stopping time T. Then  $(A_{t \wedge T})^{is}$  predictable.

*Proof.* By [DM2,VI53], the purely jump part of  $\boldsymbol{A}_t$  is

(3.9) 
$$A_{t}^{d} = \sum_{n} H_{n} 1_{\{T_{n} \leq t\}},$$

where  $H_n$  is  $\mathcal{F}_{T_n}$ -measurable and  $\{T_n\}$  are predictable stopping times with disjoint graphs. Thus  $C_t^d = \sum_n \phi(H_n) \mathbf{1}_{\{T_n \leq t\}}$  is predictable. Under the assumption in (b),  $A_{t\Lambda}^d T = A_t^d$  is predictable and  $(A_{t\Lambda}^c T)$  is continuous. Thus  $A_{t\Lambda}^c T$  is predictable.

Let  ${}^1(t < T)^{=m}t^{-A}t$  be the Doob-Meyer decomposition and suppose that  $A_t$  is continuous on  $(T, \omega)$ . By (b) in the above, we conclude  $A_t = A_{t\Lambda}T$ . Then we will say T is a quasi life time. Let  $(N_t)$  be a predictable increasing process satisfying the condition:

(3.10) 
$$N_0=1 \text{ and } \Delta N_+=0 \text{ on } [T,\infty).$$

By (b), we can assume  $N_t = N_{tA} T$  to study the process  $N_t 1_{\{t < T\}}$ . Considering  $\xi_t$  as  $1_{\{t < T\}}$ , the formula (3.3) gives us a canonical decomposition of the semimartingale  $(N_t \xi_t)$ . Because,  $N_{t-} dA_t$  and  $\xi_t dN_t = dN_t$  are predictable. Note that the canonical decomposition is unique(see [DM2]).

Proposition 3.5. Let T be a quasi-life time and  $(N_t)$  be a predictable increasing process satisfying (3.10). Then the process  $N_t 1_{\{t < T\}}$  is a supermartingale iff  $(N_t)$  satisfies the inequality

$$(3.11) dN_t/N_t \leq dA_t.$$

Then we have  $N_t \leq \tilde{N}_t$  on [0,T).

Proof. Let  $(N_t \S_t)$  be a supermartingale. Then it has a unique canonical decomposition as (local mart.)-(predictable increasing process). Thus (3.11) follows by (3.3) and the converse is clear as well. Moreover, let  $(C_t)$  be the Stieltjes logarithm of  $(N_t)$  defined by  $dC_t = dN_t/dN_t$ . Since  $dC_t \le dA_t$ , we have  $dC_t \le dA_t$  and  $\Delta C_t \le \Delta A_t$ . By the explicit expression (3.5), we conclude  $N_t \le N_t$ .

For  $0 \ll 1$ , define an increasing function  $\phi^{\alpha}$  on  $[0,\infty)$  by

(3.12) 
$$\phi^{\alpha}(\mathbf{x}) = (1-\alpha)(\mathbf{x} \wedge 1) + \alpha \frac{\mathbf{x}}{1+\mathbf{x}},$$

and a predictable increasing process  $(C_t^{\alpha})$  by (3.8) as  $\phi = \phi^{\alpha}$ . Since  $\phi^{\alpha}(x) < x \wedge I$ , we have  $dC_t^{\alpha} \le dA_t$  and  $\Delta C_t^{\alpha} < I$ . Now consider the equation

(3.13) 
$$dN_{t}^{\alpha}/N_{t}^{\alpha}=dC_{t}^{\alpha} \text{ with } N_{0}^{\alpha}=1.$$

This unique solution is

$$(3.14) N_t^{\alpha} = \exp(C_t^{\alpha})^{C} \prod_{s \leq t} (1 - \Delta C_s^{\alpha})^{-1} = \exp(A_t^{C}) \prod_{s \leq t} (1 - \phi^{\alpha} (\Delta A_s))^{-1}$$

We again consider the derivative of  $(N_t^{\alpha}\xi_t)$  by a different way from (3.3):

$$(3.15) d(N_t^{\alpha} \xi_t) = N_t^{\alpha} d\xi_t + \xi_{t-} dN_t^{\alpha} = N_t^{\alpha} dm_t - N_t^{\alpha} dA_t + \xi_{t-} dN_t^{\alpha}$$

$$= N_t^{\alpha} dm_t - (N_t^{\alpha} dA_t - dN_t^{\alpha}).$$

Since  $(N_t^{\alpha})$  is predictable, the first term of the right hand is a local martingale. By (3.13) and  $C_t^{\alpha} \leq A_t$ , the process  $(N_t^{\alpha} \leq t_t)$  is a supermartingale.

Theorem 3.6 Let  $(A_t)$  be the compensator of a quasi life time T.

- (a)  $\Delta A_t < 1$  for t < T.
- (b) Let K={ $\triangle A_T \ge 1$ }. Then  $T_K$  is a predictable stopping time.
- (c) If  $\Delta A_T$ =0, then  $(A_t)$  is continuous and T is totally inaccessible.

Proof. Let  $\tau=\inf\{t;\Delta A_t\geq 1\}$ . Then  $\tau$  is a predictable stopping time and  $\tau>0$  a.s.. Let  $N_t^0=\lim_{\alpha\to 0}N_t^\alpha$ . By virtue of the supermartingale property, we have  $E[N_t^\alpha \xi_\tau] \leq E[\xi_0]$  for  $\alpha>0$ . Therefore  $N_t^0 \xi_\tau$  is integrable. By (3.14),  $N_t^0$  is infinite on  $\{t\geq \tau\}$ , which implies  $\tau\geq T$  a.s.. Thus (a) is proved. Therefore  $T_K$  is equal to  $\tau$  which is predictable. Suppose that  $\Delta A_T=0$ . Then  $N_t^\alpha$  satisfies (3.10). By Proposition 3.5,  $N_t^\alpha \leq N_t$ . Comparing (3.5) and (3.14), since  $(1-\phi^\alpha(x))^{-1}>1+x$ , we conclude  $\Delta A_t=0$ . The proof is finished.

Remark 3.7. (a) The formula (2.12) suggests the above theorem and its inverse transformation is the form of (3.14) not (3.5). However, the equation (3.13) for

 $(A_t)$  has the solution (3.14) only if jumps of  $(A_t)$  are less than one. Clearly, (3.14) is greater than (3.5) (which implies a contradiction). This incompatibility was caused by the definition (2.10) of  $(A_t)$  in Section 2. If we defined it by

$$-dM_s/M_s=dA_s$$
 on  $[0,\infty]$ ,

it is compatible to (3.5) not to (3.14).

(b) Given a positive supermartingale  $z_t$ , we can consider a general problem: find a predictable increasing process  $N_t$  such that  $N_t z_t$  is a (local) martingale. It was discussed in Yoeurp and Meyer [YM]. Our discussion in the above is concerned with a special type of  $z_t$  and not contained in [YM].

## 4. Conservativity of (S<sub>+</sub>)

In this section, we assume that

(N) 
$$0<\xi<\omega$$
  $P_x$ -almost surely for every  $x\in E$ .

A Markov process satisfying the condition 0 < t is usually called "normal" (always assumed in this paper). If the condition t < t is not satisfied, a slight modification realizes it. Take any positive constant t and consider the process associated the semigroup  $exp(-t t)T_t$ . Clearly our problem does not change taking this process and it satisfies the condition (N). Under this condition, since j=1 is a bounded potential, we can take a predictable AF  $(A_t)$  in (3.2) which generates t (see [DM3,XV]).

Moreover, we assume that

(T) 5 is totally inaccessible,

according to the terminology due to P. A. Meyer. We recall that it means that for every increasing sequence of stopping times  $\{S_n\}$ ,

(4.1)  $\lim_{n\to\infty} S_n < \xi \quad \text{almost surely on } \{S_n < \xi \text{ for all } n \text{ and } \xi <_{\omega} \}.$ 

It is equivalent to the existence of the decomposition of  $\xi_t = I_{\{t < \xi_j\}}$  such that

(4.2) 
$$\xi_{t} = 1 + m_{t} - A_{t}$$
,

where  $m_t$  is a martingale with jumps at 5 and A is a PCAF. Also note that  $m_0 = A_0 = 0$  a.s. and

(4.3) 
$$A_{+}=A_{\zeta}$$
 on  $\{\zeta \leq t\}$  a.s..

We will assume that the "uniformity of the killing":

(U) 
$$I(t)=\inf\{P_{\mathbf{y}}[\zeta>t]; \mathbf{x} \in E\}>0 \text{ for some } t>0.$$

If the condition (U) is satisfied, then it is easy to see that it holds for every t>0, because  $P_X[S>t]=T_t 1(x)$  and I(t) is decreasing. Note that this condition fails if the process has the accessible absorbing boundary.

### Proposition 4.1. The condition

(V) 
$$\lim_{t \to 0} I(t)=1$$

implies (T) and (U).

*Proof.* Obviously, we must show (T). Let  $\{S_n\}$  be an increasing sequence of stopping times and  $S=\lim_n S_n$ . For every  $\varepsilon>0$ , we have

$$\begin{split} & P_{\mathbf{X}}[\mathbf{S} = \boldsymbol{\zeta} <_{\boldsymbol{\omega}} \text{ and } \mathbf{S}_{n} <_{\boldsymbol{\zeta}} \boldsymbol{\zeta} \text{ for all } n] \leq \lim_{n \to \boldsymbol{\omega}} P_{\mathbf{X}}[\mathbf{S}_{n} <_{\boldsymbol{\zeta}} \leq_{\mathbf{S}_{n}} +_{\boldsymbol{\varepsilon}}] = \lim_{n \to \boldsymbol{\omega}} E_{\mathbf{X}}[P_{\mathbf{X}}(\mathbf{S}_{n})[\boldsymbol{\zeta} \leq_{\boldsymbol{\varepsilon}}]] \\ & \leq \lim_{n \to \boldsymbol{\omega}} (1 - \mathbf{I}(\boldsymbol{\varepsilon})) P_{\mathbf{X}}[\mathbf{S}_{n} <_{\boldsymbol{\zeta}}] \leq (1 - \mathbf{I}(\boldsymbol{\varepsilon})) P_{\mathbf{X}}[\mathbf{S} \leq_{\boldsymbol{\zeta}}]. \end{split}$$

Taking the limit  $\varepsilon \rightarrow 0$ , we get the conclusion.

The condition (V) means that the function j is uniformly excessive and the above can be understood as a corollary of [BG,IV3.16].

In Proposition 2.7, we mentioned that the distribution of  $A_{\zeta}$  is the exponential law. We can deduce it without the help of the upper process.

Theorem 4.2. Assume that (N) and (T) hold. Then the variable  $A_{\zeta}$  has the exponential distribution with parameter one.

Proof. We show that

(4.4) 
$$E_X[A_{\zeta}^{n}]=n!$$
 for every n.

From (4.2), we have  $E_X[A_t] = P_X[S \le t]$ . By (4.3) and (N), taking the limit we obtain (4.4) for n=1. Let  $U_t = I_{\{S \le t\}} = -m_t + A_t$ . Since  $A_t$  is continuous increasing, we have

(4.5) 
$$d(A_{+}^{n})=nA_{+}^{n-1}dA_{+}=nA_{+}^{n-1}(dm_{+}+dU_{+}).$$

Therefore  $A_t^{n}=(a\ local\ martingale)+nA_{\zeta}^{n-1}U_t$ . Define

(4.6) 
$$\tau_{\varepsilon} = \inf\{t; A_{t} \ge \varepsilon\}.$$

Writing  $T=min\{t,\tau_{\varepsilon}\}$ , we have  $E_X[A_T^n]=nE_X[A_{\zeta}^{n-1}U_T]$ . Taking the limit of t and  $\varepsilon$  to  $\omega$  and by the induction, we can conclude (4.4).

By Corollary 3.2, we know that the semigroup

$$(4.7) S_t f(x) = E_x[f(X_t) \exp(A_t)].$$

is submarkov and the process  $exp(A_t)$  is a super martingale. The next lemma is fundamental in the following.

Lemma 4.3. For every function f on E and any stopping time T, we have

(4.8) 
$$\mathbb{E}_{\mathbf{X}}[\exp(\mathbf{A}_{\mathsf{T}})\mathbf{f}(\mathbf{X}_{\mathsf{T}});\mathbf{A}_{\mathsf{T}}<\varepsilon]=e^{\mathcal{E}}\,\mathbb{E}_{\mathbf{X}}[\mathbf{f}(\mathbf{X}_{\mathsf{T}});\mathbf{A}_{\mathsf{T}}<\varepsilon<\mathbf{A}_{\mathsf{\zeta}}]$$

and

$$(4.9) \qquad \qquad \mathbb{E}_{\mathbf{X}}[\exp(\mathbf{A}_{\mathbf{T}})\mathbf{f}(\mathbf{X}_{\mathbf{T}});\mathbf{A}_{\mathbf{T}} \geq \varepsilon] = e^{\varepsilon} \mathbb{E}_{\mathbf{X}}[\mathbf{S}_{\mathbf{T}-\tau}\mathbf{f}(\mathbf{X}_{\tau});\mathbf{A}_{\mathbf{T}} \geq \varepsilon]$$

where  $\tau = \tau_{\mathcal{E}}$  is the stopping time defined by (4.6). Specially, we have

(4.10) 
$$E_{\mathbf{x}}[\exp(\mathbf{A}_{\mathbf{T}})\xi_{\mathbf{T}};\mathbf{A}_{\mathbf{T}}<\varepsilon]=1-e^{\varepsilon}P_{\mathbf{x}}[\mathbf{A}_{\mathbf{T}}\geq\varepsilon].$$

Proof. The right hand side of (4.8) is

$$\exp(\varepsilon) \mathbb{E}_{\mathbf{X}}[f(\mathbf{X}_{T})P_{\mathbf{X}_{T}}[\mathbf{A}_{\zeta}>\mathbf{s}]_{\mathbf{s}=\varepsilon-\mathbf{A}_{T}};\mathbf{A}_{T}<\varepsilon].$$

Since  $P_X[A_{\zeta}>s]=e^{-s}$  by the above theorem, we get (4.8). Since  $\{A_{T}\geq \varepsilon\}=\{\tau\leq T\}$ ,

(4.9) is immediate by the strong Markov property. Let f=1. Similarly, we have

$$\begin{split} & \mathbb{E}_{\mathbf{X}}[\exp(\mathbf{A}_{\mathbf{T}}) \mathbf{S}_{\mathbf{T}}; \mathbf{A}_{\mathbf{T}} <_{\varepsilon}] \! = \! \mathbf{e}^{\varepsilon} \, \mathbb{E}_{\mathbf{X}}[\mathbf{A}_{\mathbf{T}} <_{\varepsilon} \leq \mathbf{A}_{\mathbf{S}}] \\ & = \! \mathbf{e}^{\varepsilon} \, \mathbb{P}_{\mathbf{X}}[\varepsilon \leq \mathbf{A}_{\mathbf{S}}] \! - \! \mathbf{e}^{\varepsilon} \, \mathbb{P}_{\mathbf{X}}[\mathbf{A}_{\mathbf{T}} \geq \varepsilon] \! = \! 1 \! - \! \mathbf{e}^{\varepsilon} \, \mathbb{P}_{\mathbf{X}}[\mathbf{A}_{\mathbf{T}} \geq \varepsilon]. \end{split}$$

The next proposition is a direct consequence of (4.10).

Proposition 4.4.  $(S_t)$  is conservative if and only if for some t>0,

(4.11) 
$$\lim_{\varepsilon \to \infty} e^{\varepsilon} P_{x}(A_{t} > \varepsilon) = 0 \text{ for every } x \in E.$$

We define

(4.12) 
$$J(t)=\inf\{E_{\mathbf{x}}[\exp(A_{t})\xi_{t}]; \mathbf{x} \in E\}=\inf_{\mathbf{x}} S_{t}1(\mathbf{x}).$$

We know that J(t) is decreasing and  $J(t) \ge I(t)$ .

Theorem 4.5. Assume (N), (T) and (U). Then  $(S_{+})$  is conservative.

*Proof.* Let  $\tau = \tau_{\mathcal{E}}$  be the stopping time given by (4.6). Then by (4.9)

$$\begin{split} & \mathbb{E}_{\mathbf{X}}[\exp(\mathbf{A}_{t})\mathbf{S}_{t};\mathbf{A}_{t} \succeq \boldsymbol{\varepsilon}] = \mathbf{e}^{\boldsymbol{\varepsilon}} \, \mathbb{E}_{\mathbf{X}}[\mathbf{S}_{t-\tau} \, \mathbf{1}(\mathbf{X}_{\tau}); \mathbf{A}_{t} \succeq \boldsymbol{\varepsilon}] \\ & \geq \mathbf{e}^{\boldsymbol{\varepsilon}} \, \mathbb{E}_{\mathbf{X}}[\mathbf{J}(t-\tau); \mathbf{A}_{t} \succeq \boldsymbol{\varepsilon}] \\ & \geq \mathbf{J}(t) \mathbf{e}^{\boldsymbol{\varepsilon}} \, \mathbb{P}_{\mathbf{X}}[\mathbf{A}_{t} \succeq \boldsymbol{\varepsilon}] = \mathbf{J}(t) (1 - \mathbb{E}_{\mathbf{X}}[\exp(\mathbf{A}_{t})\mathbf{S}_{t}; \mathbf{A}_{t} < \boldsymbol{\varepsilon}]) \quad (\mathbf{by} \ (4.10)) \\ & \geq 0 \end{split}$$

The first formula of this inequality goes to zero as  $\varepsilon \to \infty$ . Since  $J(t) \ge I(t) > 0$ , we conclude  $E_X[exp(A_t) \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ ]=1$ .

Corollary 4.6. Assume (N) and (V). Then  $(S_t)$  is conservative.

## 5. Case of non-conservative St

In this section, we use the same notations in Section 4 and always assume

(N) and (T).

Consider J(t) defined by (4.12). From the proof of Theorem 4.5, if J(t) is strictly positive, then we have  $(S_t)$  is conservative, and so J(t)=1. Thus we have J(t)=0 or 1 alternatively. Therefore we can study  $(S_t)$  under the assumption:

(Z) 
$$J(t)=0$$
 for some  $t>0$ .

Fix h>0. For 0 < c < 1, we define the subset  $E_c$  of E by

(5.1) 
$$E_{c} = \{x \in E; P_{x}[\zeta > h] > c\} = \{x \in E; T_{h}1(x) > c\}.$$

Under the condition (Z), since I(h)=0,  $\{E_C\}$  increases to E as c > 0. We denote the exit time from  $E_C$  by  $\sigma_C$ . Define

(5.2) 
$$S_{t}^{c}f(x)=E_{x}[exp(A_{t \wedge \sigma_{c}})f(X_{t \wedge \sigma_{c}})]$$

Since  $\sigma_{\mathcal{C}}$  is the exit time,  $A_{t \wedge \sigma_{\mathcal{C}}}$  is also a PCAF and so  $(\mathcal{S}_t^{\mathcal{C}})$  is a semigroup.

If  $\sigma_c$ =0 a.s., then  $(S_t^c)$  is trivial. However, since  $T_h 1$  is an excessive function, we can consider that it is fine continuous under a suitable assumption. For example, assume that  $(X_t)$  is a Hunt process or a right process. Then  $E_c$  is fine open, hence  $S_t^c$  is nontrivial and

(5.3) 
$$\lim_{c \to 0} \sigma_c = \xi \text{ almost surely.}$$

Theorem 5.1. Assume the condition (Z) and that  $T_h I$  is fine continuous. Let c be sufficiently small positive such that  $E_C$  is not empty. Then,  $(S_t^C)$  is a nontrivial conservative semigroup.

*Proof.* Let  $T=h \wedge \sigma_C$ . By (5.1), for every  $x \in E_C$  and s < h we have

$$S_s1(X_t) \ge S_h1(X_t) > c$$
 on  $\{t < T\}$   $P_x$ -a.s..

By the analogous way to the proof of Theorem 4.5, we have

$$\begin{split} & \mathbb{E}_{\mathbf{X}}[\exp(\mathbf{A}_{\mathbf{T}})\boldsymbol{\xi}_{\mathbf{T}}; \mathbf{A}_{\mathbf{T}} \boldsymbol{>}_{\boldsymbol{\varepsilon}}] = \mathbf{e}^{\boldsymbol{\varepsilon}} \, \mathbb{E}_{\mathbf{X}}[\mathbf{S}_{\mathbf{t}-\mathbf{\tau}} \mathbf{1}(\mathbf{X}_{\mathbf{\tau}}); \boldsymbol{\tau} \boldsymbol{<} \mathbf{T}] \\ & \geq \mathbf{e}^{\boldsymbol{\varepsilon}} \, \mathbf{c} \, \, \mathbb{P}_{\mathbf{X}}[\boldsymbol{\tau} \boldsymbol{<} \mathbf{T}] = \mathbf{c} (1 - \mathbb{E}_{\mathbf{X}}[\exp(\mathbf{A}_{\mathbf{T}})\boldsymbol{\xi}_{\mathbf{T}}; \mathbf{A}_{\mathbf{T}} \boldsymbol{\leq} \boldsymbol{\varepsilon}]) \geq 0. \end{split}$$

Since c>0, we again conclude  $E_X[exp(A_T)\$  T]=1. Thus  $S_h^C$ 1=1. By the semigroup property,  $S_t^C$ 1=1 for every t.

Remark 5.2. In the above, we gave a direct proof. However, we can get the conclusion by considering the stopped process  $(X_{t \wedge \sigma_C})$  which satisfies the condition (U).

Let  $\overline{E}=E\cup\{\Lambda\}$  and define the topology on  $\overline{E}$  by taking all sets of the form

(5.4) 
$$V_{c} = \{x \in E; P_{x}[\varsigma > h] < c\} \cup \{\Delta\}$$

as the fundamental neighbourhood of the added point  $\boldsymbol{\Delta}$  . Define

(5.5) 
$$H_{t}f(x)=E_{x}[\exp(A_{t})f(X_{t});t<\zeta]+f(\Delta)\{1-E_{x}[\exp(A_{t});t<\zeta]\} \quad \text{if } x\neq\Delta,$$
$$=f(\Delta) \quad \text{if } x=\Delta.$$

This is the simplest and usual definition of a Markov kernel for constructing a Markov process given a submarkov kernel.

We consider the example given in section 2 again. For c>0, we can take  $E_c=(c,\infty)$  essentially. Clearly, the stopped process at c has the conservative upper process. As c>0, the upper process becomes to the brownian motion on  $[0,\infty)$  which has the trap 0. In general, we can obtain the following.

Theorem 5.3. Assume the condition (Z) and that  $T_h 1$  is fine continuous. Then, for every bounded continuous function f on  $\overline{E}$ .

(5.6) 
$$\lim_{c \to 0} S_t^c f(x) = H_t f(x) \text{ for every } x \in \overline{E}.$$

*Proof.* We simply write  $\sigma$  as  $\sigma_{c}$ . Consider

$$S_{t}^{c}f(x)=E_{x}[\exp(A_{t})f(X_{t});t<\sigma,t<\xi]+E_{x}[\exp(A_{\sigma})f(X_{\sigma});\sigma\leq t,\sigma<\xi]$$

$$=I_{1}+I_{2}.$$

When c o 0,  $I_1$  converges to the first term of the right hand side of (5.5).

Moreover,

$$\begin{split} & I_2 = \mathbb{E}_{\mathbf{X}} [\exp(\mathbf{A}_{\sigma}) \{ \mathbf{f}(\mathbf{X}_{\sigma}) - \mathbf{f}(\Delta) \}; \sigma \leq \mathbf{t}, \sigma < \delta \ ] + \mathbf{f}(\Delta) \mathbb{E}_{\mathbf{X}} [\exp(\mathbf{A}_{\sigma}); \sigma \leq \mathbf{t}, \sigma < \delta \ ] \\ & = & I_3 + I_4. \end{split}$$

By the assumption, for sufficiently small c, we have  $|f(X_{\sigma})-f(\Delta)| < \varepsilon$ . Therefore

$$|I_3| \le \varepsilon E_x[\exp(A_\sigma); \sigma \le t, \sigma < \xi] \le \varepsilon$$
.

By Theorem 5.1, we have

$$1=E_{x}[\exp(A_{t});t<\sigma,t<\zeta]+E_{x}[\exp(A_{\sigma});\sigma\leq t,\sigma<\zeta].$$

Thus

$$I_{\Delta}=f(\Delta)\{1-E_{x}[\exp(A_{t});t<\sigma,t<\zeta]\},$$

which converges to the second term of (5.5).

The following proposition tells us that  $(S_t)$  can be obtained as a conditional limit. Intuitively, this fact means that the upper process is a conditional process on the set  $\{A_{\zeta} = \emptyset\}$ . However, it seems to be difficult for the author to prove it.

Theorem 5.4. For every bounded measurable function f, we have

(5.7) 
$$S_{t}f(x)=S_{t}1(x)\lim_{\varepsilon\to\infty} E_{x}[f(X_{t})|A_{t}<\varepsilon< A_{\xi}]$$

If  $(S_t)$  is conservative, then we have

(5.8) 
$$S_{t}f(x)=\lim_{\varepsilon \to \infty} E_{x}[f(X_{t})|_{\varepsilon} < A_{\xi}]$$

Proof. From (4.8), we get

$$S_t f(x) = \lim_{\varepsilon \to \infty} e^{\varepsilon} E_x [f(X_t); A_t < \varepsilon < A_{\zeta}].$$

On the other hand, we have

$$\mathrm{e}^{\mathcal{E}}\,\mathrm{P}_{\mathbf{X}}[\mathrm{A}_{\mathsf{t}}<_{\mathcal{E}}<\mathrm{A}_{\mathsf{\zeta}}] = \mathrm{e}^{\mathcal{E}}\,\mathrm{P}_{\mathbf{X}}[_{\mathcal{E}}<\mathrm{A}_{\mathsf{\zeta}}] - \mathrm{e}^{\mathcal{E}}\,\mathrm{P}_{\mathbf{X}}[_{\mathcal{E}}\leq\mathrm{A}_{\mathsf{t}}] = 1 - \mathrm{e}^{\mathcal{E}}\,\mathrm{P}_{\mathbf{X}}[_{\mathcal{E}}\leq\mathrm{A}_{\mathsf{t}}].$$

By (4.10), this formula converges to  $S_t 1(x)$  as  $\varepsilon \to \infty$ . Thus (5.7) is proved. Let

 $(S_t)$  be conservative. By Proposition 4.4, we can change the conditional form in (5.7) to that in (5.8).

Proposition 5.5. For every bounded continuous function f on E, we have

(5.9) 
$$\lim_{t\to 0} S_t f(x) = f(x) \text{ for every } x \in E.$$

*Proof.* Fix  $\varepsilon > 0$ . By Lemma 4.3, we have

$$\begin{split} & \| \mathbf{S}_{\mathbf{t}} f(\mathbf{x}) - \mathbf{e}^{\mathcal{E}} \mathbf{E}_{\mathbf{x}} [f(\mathbf{X}_{\mathbf{t}}); \mathbf{A}_{\mathbf{t}} <_{\mathcal{E}} <_{\mathbf{A}_{\mathcal{S}}}] \| = \| \mathbf{e}^{\mathcal{E}} \mathbf{E}_{\mathbf{x}} [\mathbf{S}_{\mathbf{t} - \tau} f(\mathbf{X}_{\tau}); \mathbf{A}_{\mathbf{t}} \ge_{\mathcal{E}}] \| \\ & \leq \| \mathbf{f} \| \mathbf{e}^{\mathcal{E}} \mathbf{P}_{\mathbf{x}} [\mathbf{A}_{\mathbf{t}} \ge_{\mathcal{E}}]. \end{split}$$

Since  $A_0=0$  a.s., the right hand side goes to zero as t o 0. Since f is continuous, we also have

$$\lim_{t \to 0} e^{\varepsilon} E_{x}[f(X_{t}); A_{t} < \varepsilon < A_{\zeta}] = e^{\varepsilon} P_{x}[A_{\zeta} > \varepsilon]f(x) = f(x),$$

which completes the proof.

Corollary 5.6. For every bounded continuous function f on E, we have

(5.10) 
$$\lim_{t \to 0} S_t^{C} f(x) = f(x) \text{ for every } x \in E.$$

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