POSITIVELY RECURRENT MARKOV CHAINS AND THE STEPPING STONE MODEL AS A FLEMING-VIOT PROCESS

By

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Abstract. The stepping stone model with infinitely many alleles has been studied in the framework of Fleming-Viot processes by Handa [5]. In this paper, it is investigated the strong-migration limit of the average number of distinct elements in a sample of finite particles in the stationary state of the model, where the results in [5] are applied. To obtain our result, we investigate a problem on positively recurrent Markov chains, and make use of the results in Shiga, Shimizu and Soshi [12].

1. Introduction

We will discuss the stepping stone model as a Fleming-Viot process, and the related Markov chains determined by the migration rates. Fleming-Viot processes have been investigated by many authors. See Ethier and Griffiths [1], Ethier and Kurtz [2] and [3]. One of important applications of Fleming-Viot processes is to formulate mathematically the infinite allele model in population genetics which was given by Kimura and Crow [7]. The Fleming-Viot process describing the stepping stone model with infinitely many alleles has been studied by Handa [5]

First, we explain the Fleming-Viot process describing the stepping stone model, and our problem on the measure-valued diffusion. Let E be the 1-dimensional interval [0, 1], and S be a countable (or finite) set. The set P(E) of probability measures on E is endowed with the weak topology. Define \mathcal{P}

$$\mathcal{P} = P(E)^S = \{ \widetilde{\mu} = \{ \mu_k \}_{k \in S}; \mu_k \in P(E), k \in S \},$$

with the product topology. In the following, for a topological space X, B(X) stands for the space of bounded Borel measurable functions on X. For each positive integer k, X^k denotes the k-fold product of X.

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Let N be the set of positive integers, and θ be a positive number. For each $n \in N$ and any $f \in B(E^n)$, define $L^{(n)}f$ by

$$(L^{(n)}f)(x_1,x_2,\ldots,x_n)=\sum_{i=1}^n(L_if)(x_1,x_2,\ldots,x_n),$$

where

$$(L_i f)(x_1, x_2, \ldots, x_n) = \frac{\theta}{2} \left(\int_0^1 f(x_1, \ldots, x_{i-1}, u, x_{i+1}, \ldots, x_n) du - f(x_1, \ldots, x_n) \right).$$

For $n \in \mathbb{N}$, $f \in B(E^n)$ and $1 \le i < j \le n$, $\Phi_{ij}^{(n)} f$ is defined by

$$(\mathbf{\Phi}_{ij}^{(n)}f)(x_1,x_2,\ldots,x_{n-1})=f(x_1,\ldots,x_{j-1},x_i,x_j,x_{j+1},\ldots,x_{n-1}).$$

Here, in the right-hand side of the above equality, the j-th variable x_j is replace by x_i , and the variables $x_{j+1}, x_{j+2}, \ldots, x_n$ are replaced by $x_j, x_{j+1}, \ldots, x_{n-1}$ respectively.

Let us introduce a matrix $\{m_{k'k}\}_{k',k\in S}$, which describes migration rates. Suppose

$$m_{k'k} \geq 0 \quad ext{if} \quad k
eq k' \quad ext{and} \quad m_{kk} = -\sum_{k':k'
eq k} m_{k'k}.$$

Furthermore, we assume that $\sup_{k \in S} |m_{kk}| < +\infty$. For $k = (k_1, k_2, \dots, k_n) \in S^n$, $\beta_j k$ and $\gamma_i(k')k$ are defined by

$$\beta_j \mathbf{k} = (k_1, k_2, \dots, k_{j-1}, k_{j+1}, \dots, d_n) \in S^{n-1},$$

and

$$\gamma_i(k')k = (k_1,\ldots,k_{i-1},k',k_{i+1},\ldots,k_n) \in S^n.$$

Namely, β_j eliminates the *j*-th coordinate of k, and $\gamma_i(k')$ replaces the *i*-th element of k by k'. For $k = (k_1, k_2, \ldots, k_n) \in S^n$, $\overline{\mu}_k$ denotes $\mu_{k_1} \times \mu_{k_2} \times \cdots \times \mu_{k_n}$ (direct product), and $\phi_{f,k}(\overline{\mu})$ is defined by $\phi_{f,k}(\overline{\mu}) = \langle f, \overline{\mu}_k \rangle$ for $f \in B(E^n)$.

Now, we are in a position to define the operator \mathcal{L} by

$$(\mathcal{L}\phi_{f,\boldsymbol{k}})(\overline{\mu}) = \sum_{1 \leq i < j \leq n, k_i = k_j} (\langle \Phi_{ij}^{(n)} f, \overline{\mu}_{\beta_j \boldsymbol{k}} \rangle - \langle f, \overline{\mu}_{\boldsymbol{k}} \rangle)$$

$$+ \langle L^{(n)} f, \overline{\mu}_{\boldsymbol{k}} \rangle + \sum_{i=1}^n \sum_{\boldsymbol{k}' \in S} m_{\boldsymbol{k}' k_i} \langle f, \overline{\mu}_{\gamma_i(\boldsymbol{k}') \boldsymbol{k}} \rangle.$$
 (1.1)

It is shown in Handa [5] that the martingale problem for the operator \mathcal{L} is well posed, and that the \mathcal{P} -valued diffusion $\{\mu(t)\}$ determined by \mathcal{L} has a unique stationary distribution, which is denoted by $\overline{Q}(d\overline{\mu})$.

Define

$$I_n = \{\alpha = \{\alpha_k\}_{k \in S}; \alpha_k \in Z_+, \sum_{k \in S} \alpha_k = n\}, \text{ and } |\alpha| = \sum_{k \in S} \alpha_k.$$

where Z_{+} denotes the set of non-negative integers, and set

$$I=\bigcup_{n=1}^{+\infty}I_n.$$

Consider the mapping $\Psi: \bigcup_{n=1}^{+\infty} S^n \to I$, defined by

$$\Psi(\mathbf{k}) = \{\alpha_{\mathbf{k}}\}_{\mathbf{k} \in S}, \mathbf{k} \in S^n,$$

where α_k = the cardinality of the set $\{i; k_i = k\}$. Let $\overline{\mu}_{\alpha} = \prod_{k \in S, \alpha_k > 0} \mu_k^{\alpha_k}$, where $\mu_k^{\alpha_k}$ denotes the α_k -fold product of μ_k . If a function $f \in B(E^n)$ is symmetric, and $\Psi(k) = \Psi(k') = \alpha \in I_n$, then $\langle f, \overline{\mu}_{\alpha} \rangle$ is well-defined and we have

$$\langle f, \widetilde{\mu}_{\mathbf{k}} \rangle = \langle f, \widetilde{\mu}_{\mathbf{k'}} \rangle = \langle f, \widetilde{\mu}_{\alpha} \rangle.$$

Note that for a symmetric function $f \in B(E^n)$, $\Phi_{ij}^{(n)} f$ and $L^{(n)} f$ are symmetric in $B(E^{n-1})$ and $B(E^n)$ respectively. In the following, we assume that the migration rates $m_{k'k}$, k', $k \in S$ are written in the form

$$m_{k'k} = mr_{k'k}, \quad k', k \in S,$$

with a positive constant m. The stationary distribution $\widetilde{Q}(d\widetilde{\mu})$ is denoted by $\widetilde{Q}_m(d\widetilde{\mu})$, since it depends on m. Define $f_n \in B(E^n)$ by $f_n(x_1, x_2, \ldots, x_n) =$ the number of distinct elements of the sequence (x_1, x_2, \ldots, x_n) . Note that the function f_n is symmetric.

Our problem on the Fleming-Viot process is to investigate the asymptotic behavior of the integral

$$\mathcal{A}_m(lpha) := \int \mathcal{F}_lpha(\widetilde{\mu}) \widetilde{Q}_m(d\widetilde{\mu}), \qquad \qquad (1.2)$$

as $m \to \infty$, where

$$\mathcal{F}_{\alpha}(\overline{\mu}) = \langle f_n, \widetilde{\mu}_{\alpha} \rangle, \alpha \in I_n.$$

Here, we explain what the quantity $\mathcal{A}_m(\alpha)$ means. Choose n genes from the colonies according to $\alpha \in I_n$. Namely, α_k genes are chosen from colony k, for each $k \in S$. $\mathcal{F}_{\alpha}(\overline{\mu})$ means the expectation of the number of distinct elements in a sample of n genes with respect to the measure $\overline{\mu} \in \mathcal{P}$. Since $\mathcal{A}_m(\alpha)$ is the

average of $\mathcal{F}_{\alpha}(\overline{\mu})$ with respect to $\overline{Q}_{m}(d\overline{\mu})$, $\mathcal{A}_{m}(\alpha)$, $\alpha \in I_{n}$, should be called the average expectation of the number of alleles in a sample of n genes.

The asymptotic behavior of a quantity for stepping stone models, as $m \to \infty$, is called that of the strong-migration limit. The strong-migration limit in geographically structured population was investigated in Nagylaki [8], Notohara [9] and [10]. Notohara [10] discussed the same problem as ours. In [10], it is assumed that the migration process is a random walk on the d-dimensional torus S with lattice points. Under this assumption, he got fruitful results on the strong-migration limit of the stepping stone model. In this paper, the set S is assumed to be a countable set, so that it will be assumed the finiteness of the moments of the first returning time for the migration process to investigate the asymptotic behavior of $\mathcal{A}_m(\alpha)$ in the strong-migration limit.

Let us introduce three kinds of Markov chains. Define $q_{kk'}, k, k' \in S$ by

$$q_{kk'}=r_{k'k}.$$

Let $q_k = -q_{kk}$. The minimal Markov chain generated by $\{q_{kk'}\}_{k,k'\in S}$ is denoted by $\{x(t), P_k\}$. It is conservative since $\{q_k\}_{k\in S}$ is bounded by the assumption on the migration rates. In the following, the Markov chain $\{x(t), P_k\}$ is sometimes called the 1-particle system. We consider the n-fold direct product of the 1-particle system $\{x(t), P_k\}$, which may be called the n-particle system. The n-particle system, whose state space is S^n , is denoted by $\{x(t)\}_{t\geq 0}$. Define the matrix $\tilde{q}_{\alpha\beta}$ on I_n , by

$$\widetilde{q}_{\alpha\beta} = \alpha_k r_{k'k} \qquad \text{if} \quad \beta = \alpha - \epsilon_k + \epsilon_{k'},$$

$$= \sum_{k} \alpha_k r_{kk} \qquad \text{if} \quad \beta = \alpha,$$

$$= 0 \qquad \text{otherwise,}$$
(1.3)

where $\epsilon_k = \{\delta_{k,l}\}_{l \in S} \in I_1$. The Markov chain with state space I_n , generated by $\{\overline{q}_{\alpha\beta}\}_{\alpha,\beta\in I_n}$, is denoted by $\{\overline{\alpha}(t)\}_{t\geq 0}$. We see that

$$\left\{\Psi(\boldsymbol{x}(t))\right\}_{t\geq0}\stackrel{\mathrm{law}}{=}\left\{\widetilde{\alpha}(t)\right\}_{t\geq0}.$$

In this paper, we assume the following.

ASSUMPTION 1. The Markov chain $\{x(t), P_k\}$, the 1-particle system, is irreducible and positively recurrent, whose stationary distribution is denoted by ν .

Under this assumption, the *n*-particle system $\{x(t)\}_{t\geq 0}$ and the Markov chain $\{\alpha(t)\}_{t\geq 0}$, mapped by Ψ from the *n*-particle system are both positively recurrent,

whose stationary distributions are $\nu_n(k) = \prod_{i=1}^n \nu(k_i)$, for $k = (k_1, k_2, \dots, k_n) \in S^n$, and $\widetilde{\nu}_n = \frac{n!}{\prod_k \alpha_k!} \nu(k)^{\alpha_k}$ for $\alpha = (\alpha_k)_{k \in S}$ respectively.

Second, we explain our problem on the 1-particle system $\{x(t), P_k\}$. Let us consider the equation on a function u_m

$$m\sum_{k'}q_{kk'}u_m(k') - V(k)u_m(k) = -f(k), \qquad k \in S,$$
 (1.4)

with a positive constant m and $f \in B(S)$, where V(k) satisfies $c_1 \leq V(k) \leq c_2$ for some positive constants c_1, c_2 . Recall that the equation (1.4) has a unique bounded solution u_m if and only if the Markov chain $\{x(t), P_k\}$ is conservative (Feller [4], Shiga [11]), and it is easy to see that

$$\lim_{m\to\infty}u_m(k)=\langle f\rangle/\langle V\rangle,$$

where $\langle g \rangle = \sum_k g(k)\nu(k)$ for $g \in B(S)$. Our problem is the following. Under what condition does the sequence $m(u_m(k) - \langle f \rangle / \langle V \rangle)$ have a finite limit as m tends to infinity? Our result on the problem mentioned above of the Markov chain $\{x(t), P_k\}$ is formulated as the next theorem. The expectation with respect to P_k is denoted by $E_k[\cdot]$. Define \mathcal{V} , a class of bounded functions on S, by

$$V = \{V \in B(S); \text{ there exist positive constants } c_1 \text{ and } c_2 \text{ such that } c_1 \leq V(k) \leq c_2, \text{ for any } k \in S\}.$$

Let T be the first returning time, namely

$$T = \inf\{t > 0; x(t) = x(0) \text{ and } x(s) \neq x(0) \text{ for some } s < t\}.$$

THEOREM 1. There exists a constant c depending on k, V, and f for any $k \in S$, $V \in \mathcal{V}$ and $f \in B(S)$ such that

$$\lim_{m\to\infty} m(u_m(k) - \langle f \rangle / \langle V \rangle) = c,$$

if and only if

$$E_{k_0}[T^2] < \infty \quad for \ some \quad k_0 \in S.$$
 (1.5)

Remark. Note that the condition $E_{k_0}[T^2] < \infty$ for some $k_0 \in S$ implies $E_k[T^2] < \infty$ for all $k \in S$. Since the matrix $\{q_{kk'}\}$ is defined at the beginning by the migration rates $\{m_{k'k}\}$ of the stepping stone model, it is assumed in Theorem 1 that q_k is bounded in $k \in S$. However, it suffices to assume on $\{q_{kk'}\}$ that the minimal Markov chain generated by $\{q_{kk'}\}$ is conservative, in order to prove Theorem 1 disregarding the stepping stone model.

Here, we apply this result to the problem of the Fleming-Viot process describing a stepping stone model. It is easy to see by using (3.3) and (3.4) below that, under Assumption 1, the equality

$$\lim_{m \to +\infty} \mathcal{A}_m(\alpha) = 1 + \theta \sum_{l=2}^n \frac{l}{\overline{\nu}_l + (\theta - 1)l} = 1 + \theta \sum_{l=2}^n \frac{1}{(l-1)\sum_k \nu(k)^2 + \theta}$$

holds for $\alpha \in I_n$, $n \geq 2$, where $\tilde{\nu}_l = \sum_k \sum_{\alpha \in I_l} \alpha_k^2 \tilde{\nu}_l(\alpha) = (l^2 - l) \sum_k \nu(k)^2 + l$. The right-hand side of the above equality is denoted by $\mathcal{A}_{\infty}(\alpha)$. Recall that the average number of alleles in randomly chosen n genes for the ordinary infinite allele model is known to be equal to

$$1+\theta\sum_{l=2}^n\frac{1}{l-1+\theta}.$$

See Hartl and Clark [6]. By Shiga, Shimizu and Soshi [12], we see that the finiteness of the expectation of T^2 for the 1-particle system implies that of the second moment of the first returning time for the *n*-particle system. Therefore, under the assumption that $E_k[T^2] < \infty$ for some $k \in S$, the first returning time for the Markov chain $\tilde{\alpha}(t)$, mapped by Ψ from the *n*-particle system has the finite second moment. Making use of Theorem 1 and the results of Shiga, Shimizu and Soshi [12], we can obtain the next theorem.

THEOREM 2. Assume for the 1-particle system $\{x(t), P_k\}$ that $E_k[T^2] < \infty$ for some k. Then, there exists a constant c' depending on α for each $\alpha \in I_2$ such that

$$\lim_{m\to+\infty} m\{\mathcal{A}_m(\alpha) - \mathcal{A}_\infty(\alpha)\} = c'.$$

2. Proof of Theorem 1

By the Feynman-Kac formula, we see that the solution $u_m(k)$ of (1.4) can be written as follows (Shiga [11]):

$$u_m(k) = \frac{1}{m} \int_0^{+\infty} E_k \left[\exp\left\{ -\frac{1}{m} \int_0^s V(x(u)) du \right\} f(x(s)) \right] ds \qquad (2.1)$$
$$= \int_0^{+\infty} E_k \left[\exp\left\{ -\frac{1}{m} \int_0^{mt} V(x(u)) du \right\} f(x(mt)) \right] dt. \qquad (2.2)$$

The ergodic property of the 1-particle system and (2.2) imply that

$$\lim_{m\to\infty} u_m(k) = \langle f \rangle / \langle V \rangle.$$

Making use of the strong Markov property of the 1-particle system and (2.1), we obtain

$$u_{m}(k) = \frac{E_{k} \left[\int_{0}^{T} \exp\left\{ -\frac{1}{m} \int_{0}^{t} V(x(u)) du \right\} f(x(t)) dt \right]}{m \left(1 - E_{k} \left[\exp\left\{ -\frac{1}{m} \int_{0}^{T} V(x(u)) du \right\} \right] \right)}.$$
 (2.3)

Define $\phi(z)$, $\psi(z)$ and v(z) by

$$\phi(z) = E_k \left[\int_0^T \exp\left\{-z \int_0^t V(x(u)) du \right\} f(x(t)) dt \right],$$
 $\psi(z) = \frac{1}{z} \left\{ 1 - E_k \left[\exp\left\{-z \int_0^T V(x(u)) du \right\} \right] \right\} \equiv \frac{1}{z} \left\{ 1 - \eta(z) \right\},$

and

$$v(z) = rac{\phi(z)}{\psi(z)}$$

for a positive number z, respectively. Note that $\phi(0+) = E_k[T]\langle f \rangle$, $\psi(0+) = E_k[T]\langle V \rangle$, and that $v(0) = \langle f \rangle / \langle V \rangle$. First, we assume that $E_{k_0}[T^2] < \infty$ for some $k_0 \in S$. In order to show that the condition (1.5) is sufficient, it suffices to prove that v(z) is right-differentiable at z=0. Since $E_k[T^2] < +\infty$ for any $k \in S$, as mentioned in Remark of Section 1, we see that

$$\lim_{z\downarrow 0}\phi'(z)=-E_k\bigg[\int_0^T\!\!\int_0^t V(x(u))duf(x(t))dt\bigg].$$

By a simple calculation, we see

$$\psi'(z) = \frac{\eta(z) - 1 - z\eta'(z)}{z^2}.$$

Noting that $E_k[T^2] < +\infty$ implies the existence of $\eta''(0+)$, we get

$$\lim_{z\downarrow 0}\psi'(z)=-\frac{1}{2}\eta''(0+)=-\frac{1}{2}E_k\bigg[\bigg(\int_0^TV(x(u))du\bigg)^2\bigg].$$

Hence, we conclude that v'(0+) exists, namely,

$$v'(0+) = \frac{\phi'(0+)\psi(0+) - \phi(0+)\psi'(0+)}{\psi(0+)^2}.$$

To complete the proof of Theorem 1, it is sufficient to show the next lemma.

LEMMA 2.1. Let $f(k) \equiv 1$, $V(k) = I_{\{k_0\}}(k) + 1$. Then, the statement that

$$u_m(k_0)-rac{1}{1+
u(k_0)}=O(1/m),\quad m o +\infty,$$

holds if and only if $E_{k_0}[T^2] < +\infty$.

Proof. Noting that $f(k) \equiv 1$, we have for the initial state k_0

$$\phi(z) = E_{k_0} \left[\int_0^T \exp \left\{ -z \int_0^t V(x(u)) du \right\} dt \right].$$

Let ζ be the first jumping time, and T_0 be the first passage time to the state k_0 , then

$$\begin{aligned} \phi(z) &= E_{k_0} \left[\int_0^{\zeta} \exp\left\{ -z \int_0^t V(x(u)) du \right\} dt \right] + E_{k_0} \left[\int_{\zeta}^T \exp\left\{ -z \int_0^t V(x(u)) du \right\} dt \right] \\ &= E_{k_0} \left[\int_0^{\zeta} \exp\left\{ -2zt \right\} dt \right] + E_{k_0} \left[\int_{\zeta}^T \exp\left\{ -z(2\zeta + t - \zeta) \right\} dt \right] \\ &= E_{k_0} \left[\frac{1}{2z} \left(1 - e^{-2z\zeta} \right) \right] + E_{k_0} \left[e^{-2z\zeta} E_{x_{\zeta}} \left[\int_0^{T_0} \exp(-zu) du \right] \right] \\ &= \frac{1}{q_{k_0} + 2z} + \frac{q_{k_0}}{z(q_{k_0} + 2z)} \left(1 - E_{k_0} \left[E_{x_{\zeta}} \left[e^{-zT_0} \right] \right] \right) \\ &= \frac{1}{z(q_{k_0} + 2z)} \left(z + q_{k_0} - q_{k_0} E_{k_0} \left[E_{x_{\zeta}} \left[e^{-zT_0} \right] \right] \right). \end{aligned}$$

Similarly, we see for the initial state k_0 that

$$\psi(z) = \frac{1}{z(q_{k_0} + 2z)} (2z + q_{k_0} - q_{k_0} E_{k_0} [E_{x_{\zeta}}[e^{-zT_0}]]).$$

Hence, we obtain

$$v(z) = 1 - \frac{1}{q_{k_0} + 2z} \frac{1}{\psi(z)},$$

so that v(z) is right-differentiable at z=0 if and only if $\frac{\psi(z)-\psi(0+)}{z}$ has a finite limit as $z\downarrow 0$. Since $\psi(0+)=E_{k_0}\left[\int_0^T V(x(u))du\right]$, we see that

$$\frac{\psi(z) - \psi(0+)}{z} = \frac{1}{z^2} \left(1 - E_{k_0} \left[z \int_0^T V(x(u)) du \right] - E_{k_0} \left[\exp \left\{ -z \int_0^T V(x(u)) du \right\} \right] \right)
= (-1) E_{k_0} \left[\int_0^{\int_0^T V(x(u)) du} dt \int_0^t e^{-zs} ds \right].$$

Consequently,

$$\lim_{z\downarrow 0}\frac{\psi(z)-\psi(0+)}{z}=(-1)\frac{1}{2}E_{k_0}\left[\left(\int_0^T V(x(u))du\right)^2\right].$$

Noting that

$$T^2 \leq \left(\int_0^T V(x(u))du\right)^2 \leq 4T^2,$$

we can conclude that there exists a finite limit $\lim_{z\downarrow 0} \frac{\psi(z)-\psi(0+)}{z}$ if and only if $E_{k_0}[T^2]<+\infty$. Thus, the proof of Lemma 2.1 is complete.

3. Proof of Theorem 2

First, let us introduce another infinitesimal matrix $q_{\alpha\beta}^m$ on I defined by

$$\begin{split} q^m_{\alpha\beta} &= \binom{\alpha_k}{2} + \frac{\theta}{2} \alpha_k & \text{if} \quad \beta = \alpha - \epsilon_k \quad \text{and} \quad |\alpha| \geq 2 \\ &= \alpha_k m r_{k'k} & \text{if} \quad \beta = \alpha - \epsilon_k + \epsilon_{k'}, \ k \neq k' \\ &= -\sum_{k \in S} \left(\binom{\alpha_k}{2} + \frac{\theta}{2} \alpha_k \right) I_{\{|\alpha| \geq 2\}}(\alpha) + \sum_{k \in S} \alpha_k m r_{kk} & \text{if} \quad \beta = \alpha \\ &= 0 & \text{otherwise,} \end{split}$$

where $\alpha + \beta$ and $\alpha - \beta$ are defined componentwisely for $\alpha, \beta \in I$. Here, $\binom{n}{k}$ means 0 if n < k.

Note that the equalities

$$(L_i f_n)(x_1, x_2, \ldots, x_n) = rac{ heta}{2}(1 + f_{n-1}(x_1, \ldots, x_{i-1}, x_{i+1}, \ldots, x_n) - f_n(x_1, x_2, \ldots, x_n)),$$

and

$$\begin{split} (\mathcal{LF}_{\alpha})(\widetilde{\mu}) &= \sum_{k \in S} \binom{\alpha_k}{2} \left(\langle f_{n-1}, \widetilde{\mu}_{\alpha - \epsilon_k} \rangle - \langle f_n, \widetilde{\mu}_{\alpha} \rangle \right) + \frac{\theta}{2} |\alpha| \\ &+ \sum_{k \in S} \frac{\theta}{2} \left(\langle f_{n-1}, \widetilde{\mu}_{\alpha - \epsilon_k} \rangle - \langle f_n, \widetilde{\mu}_{\alpha} \rangle \right) \\ &+ \sum_{k \in S} \sum_{k' \in S} \alpha_k m_{k'k} \langle f_n, \widetilde{\mu}_{\alpha - \epsilon_k + \epsilon_{k'}} \rangle \quad \text{ for } \quad |\alpha| = n \geq 2. \end{split}$$

hold. Since

$$\int (\mathcal{LF}_{m{lpha}})(\overline{\mu}) ar{Q}_{m{m}}(d\overline{\mu}) = 0,$$

we obtain

$$\sum_{\beta} q_{\alpha\beta}^{m} \mathcal{A}_{m}(\beta) + \frac{\theta}{2} |\alpha| = 0 \quad \text{if} \quad |\alpha| \ge 2, \tag{3.1}$$

$$\mathcal{A}_m(\alpha) = 1 \quad \text{if} \quad |\alpha| = 1. \tag{3.2}$$

Let $(\alpha(t), P_{\alpha}^{m})$, $t \geq 0$, $\alpha \in I$, be the Markov chain generated by $\{q_{\alpha\beta}^{m}\}_{\alpha,\beta \in I}$. E_{α}^{m} denotes the expectation with respect to P_{α}^{m} . Define \mathcal{T} and \mathcal{T}_{1} by

$$\mathcal{T}=\inf\{t\geq 0; |\alpha(t)|=1\},\,$$

and

$$\mathcal{T}_1=\inf\{t\geq 0; |\alpha(t)|<|\alpha(0)|\},$$

where inf $\emptyset = +\infty$. By (3.1) and (3.2), we obtain for $\alpha \in I_n$, $n \ge 2$,

$$\mathcal{A}_{m}(\alpha) = 1 + E_{\alpha}^{m} \left[\int_{0}^{T} \frac{\theta}{2} |\alpha(s)| ds \right]$$

$$= \frac{n\theta}{2} E_{\alpha}^{m} [\mathcal{T}_{1}] + E_{\alpha}^{m} [\mathcal{A}_{m}(\alpha(\mathcal{T}_{1}))]. \tag{3.3}$$

Observe that $\mathcal{A}_m^1(\alpha) := E_{\alpha}^m[\mathcal{T}_1]$ satisfies

$$\sum_{\beta \in I_n} \widetilde{q}_{\alpha\beta} \mathcal{A}_m^1(\beta) - \frac{1}{m} \frac{1}{2} \Big(\sum_i \alpha_i^2 + (\theta - 1)n \Big) \mathcal{A}_m^1(\alpha) = -\frac{1}{m}, \tag{3.4}$$

with the matrix $\{\widetilde{q}_{\alpha\beta}\}_{\alpha,\beta\in I_n}$ defined by (1.3). Since the minimal Markov chain $\{\widetilde{\alpha}(t),P_{\alpha}\}$ on I_n generated by the matrix $\{\widetilde{q}_{\alpha\beta}\}_{\alpha,\beta\in I_n}$ is conservative and positively recurrent, as explained in Section 1. Therefore, we can apply Theorem 1 to $\{\widetilde{\alpha}(t),P_{\alpha}\}$. Let T be the first returning time of $\{\widetilde{\alpha}(t),P_{\alpha}\}$, then by the results of Shiga, Shimizu and Soshi [12], as mentioned in Section 1, we see that $E_{\alpha}[T^2] < +\infty$, where $E_{\alpha}[\cdot]$ denotes the expectation with respect to P_{α} . Thus, we obtain the next lemma.

LEMMA 3.1. For each $\alpha \in I_n$, there exists a finite limit $\lim_{m\to\infty} m(\mathcal{A}_m^1(\alpha) - \frac{2}{\overline{\nu_n} + (\theta-1)n})$.

If $\alpha \in I_2$, the second term of the right-hand side of (3.3) is equal to 1, so that Lemma 3.1 proves Theorem 2.

4. An Example

Here, we present a Markov chain for which one can get the critical parameter for the second moment of the first returning time T to be finite. Furthermore, for the example, it is made clear the speed of convergence of the solution u_m for the equation (1.4) with specified functions V and f, when the second moment of T is infinite. Proposition 4.1 and Lemma 4.2, given below, are essentially due to Professor T. Shiga.

Consider a continuous time minimal Markov chain $\{x(t), P_i\}$ in $S = Z_+$ which is governed by the following infinitesimal matrix $Q = \{q_{i,j}\}$.

$$q_{i,j} = egin{cases} 1 & ext{ (if } j=i+1) \ lpha_1 & ext{ (if } j=0) \ -(1+lpha_i) & ext{ (if } j=i>0), \ q_{0,j} = \delta_{1,j} & ext{ } (j\geq 1) & q_{0,0} = -1, \ q_{i,j} = 0 & ext{ for any other } i,j\in S. \end{cases}$$

and

Recall that T denotes the first returning time of the continuous time Markov chain $\{x(t), P_i\}$. Then it is easy to show that

$$E_0(e^{-\lambda T}) = \sum_{i=1}^{\infty} \frac{\alpha_i}{1+\lambda} \prod_{k=1}^{i} \frac{1}{1+\lambda+\alpha_k}, \qquad (4.1)$$

from which it follows that $\{x(t), P_i\}$ is recurrent if and only if

$$\sum_{i=1}^{\infty} \alpha_i = \infty. \tag{4.2}$$

Now we specialize $\{\alpha_i\}$ as follows:

$$\alpha_i = \frac{\kappa}{i} \quad (i \ge 1), \tag{4.3}$$

with a constant $\kappa > 0$. Inserting this to (4.1) we see

$$E_0(e^{-\lambda T}) = \sum_{n=1}^{\infty} \frac{\kappa \Gamma(n) \Gamma\left(\frac{\kappa}{1+\lambda} + 1\right)}{(1+\lambda)^{n+1} \Gamma\left(\frac{\kappa}{1+\lambda} + n + 1\right)}$$
$$= \sum_{n=1}^{\infty} \frac{\kappa}{(1+\lambda)^{n+1}} B\left(n, \frac{\kappa}{1+\lambda} + 1\right),$$

where $B(\alpha, \beta)$ denotes the Beta function, so that it holds

$$E_0(e^{-\lambda T}) = \frac{\kappa}{1+\lambda} \int_0^1 \frac{t^{\frac{\kappa}{1+\lambda}}}{\lambda+t} dt. \tag{4.4}$$

Proposition 4.1.

$$E_0(T^{\alpha}) < \infty$$
 if and only if $\alpha < \kappa$.

Proof. Using Taylor's expantion of $1/(\lambda + t)$, from (4.4) one gets

$$E_0(e^{-\lambda T}) = \kappa \sum_{m=0}^n \frac{(-\lambda)^m}{\kappa - m(\lambda + 1)} + \frac{\kappa(-\lambda)^{n+1}}{\lambda + 1} \int_0^1 \frac{t^{\frac{\kappa}{\lambda + 1} - n - 1}}{\lambda + t} dt$$
$$= \kappa \sum_{m=0}^n \frac{(-\lambda)^m}{\kappa - m(\lambda + 1)} + \frac{\kappa(-1)^{n+1}}{\lambda + 1} \lambda^{\kappa/(\lambda + 1)} \int_0^{1/\lambda} \frac{t^{\frac{\kappa}{\lambda + 1} - n - 1}}{1 + t} dt.$$

Let $n < \kappa \le n+1$ $(n \in \mathbb{Z}_+)$. Denoting the last term by $b(\lambda)$, we obtain that if $n < \kappa < n+1$,

$$\lim_{\lambda \searrow 0} (-1)^{n+1} \frac{b(\lambda)}{\lambda^{\kappa}} = \kappa \int_0^\infty \frac{t^{\kappa - n - 1}}{1 + t} dt, \tag{4.5}$$

and if $\kappa = n + 1$,

$$b(\lambda) \sim \kappa(-\lambda)^{n+1} \log \frac{1}{\lambda} \quad (\lambda \searrow 0).$$
 (4.6)

Therefore, by applying the following lemma, the proof of Proposition 4.1 is completed.

LEMMA 4.1. For a nonnegative random variable Y, let

$$\varphi(\lambda) = E(e^{-\lambda Y}) \quad (\lambda > 0).$$

Suppose that $\varphi(\lambda)$ has the form,

$$\varphi(\lambda) = \sum_{m=0}^{n} a_m \lambda^m + b(\lambda),$$

where for a $\kappa \in (n, n+1]$, either

$$b(\lambda) \sim \beta_0 \lambda^{\kappa}, \quad as \quad \lambda \downarrow 0,$$

with $\beta_0(\neq 0) \in R$ or

$$\lim_{\lambda \searrow 0} (-1)^{n+1} \frac{b(\lambda)}{\lambda^{\kappa}} = \infty$$

and for every $\beta < \kappa$

$$\lim_{\lambda \searrow 0} (-1)^{n+1} \frac{b(\lambda)}{\lambda^{\beta}} = 0.$$

Then it holds that

$$E(Y^{\alpha}) < \infty$$
 for $\alpha < \kappa$, and $E(Y^{\kappa}) = \infty$.

Proof of Lemma 4.2 is standard, so it is omitted.

We see by Proposition 4.1 that $\{x(t), P_i\}$ is positively recurrent if and only if $\kappa > 1$, so that we investigate the case of $\kappa > 1$ in the following. Let us consider (1.4) for this example, with specified functions V and f. Let $V(i) = 1 + I_{\{0\}}(i)$, $f(i) \equiv 1$, where $I_{\{0\}}$ is the indicator function on the set $\{0\}$. Namely, we consider the equation

$$m\sum_{j}q_{ij}u_{m}(j)-V(i)u_{m}(i)=-1, (4.7)$$

with a positive constant m. The unique solution u_m of (4.7) satisfies

$$\lim_{m\to+\infty}u_m(0)=\frac{1}{1+\nu(0)},$$

with $\nu(0) = 1 - \frac{1}{\kappa}$, and, by the same argument as the proof of Lemma 2.1, it is shown that

$$u_m(0) = 1 - \frac{1}{1 + 2z} \frac{1}{\psi(z)},\tag{4.8}$$

where z = 1/m, and

$$\psi(z) = \frac{1}{z(1+2z)} (2z + 1 - E_1[e^{-zT_0}]), \tag{4.9}$$

where T_0 denotes the first passage time to the state 0. Note that

$$E_1[e^{-zT_0}] = \kappa \int_0^1 t^{\frac{\kappa}{1+z}} \frac{1}{z+t} dt. \tag{4.10}$$

Then, we obtain the next statement.

Proposition 4.2. If $\kappa > 2$,

$$u_m(0)-rac{1}{1+
u(0)}\sim -k_1(\kappa)rac{1}{m},\ m o +\infty,$$

if $\kappa=2$,

$$u_m(0) - \frac{1}{1+\nu(0)} \sim -\frac{2}{9} \frac{1}{m} \log m, \ m \to +\infty,$$

and if $1 < \kappa < 2$,

$$u_m(0) - \frac{1}{1 + \nu(0)} \sim -k_2(\kappa) \left(\frac{1}{m}\right)^{\kappa - 1}, \ m \to +\infty,$$

where

$$k_i(\kappa)>0, \ (i=1,2), \ \lim_{\kappa\downarrow 2}k_1(\kappa)=+\infty, \ \lim_{\kappa\uparrow 2}k_2(\kappa)=+\infty,$$

and $\nu(0) = 1 - 1/\kappa$.

Proof. Recall the equality in the proof of Proposition 4.1,

$$E_0[e^{-zT}] = \kappa \sum_{l=0}^{n} \frac{(-z)^l}{\kappa - l(z+1)} + b(z), \tag{4.11}$$

where

$$b(z) = \frac{\kappa(-1)^{n+1}}{z+1} z^{\frac{\kappa}{z+1}} \int_0^{\frac{1}{z}} \frac{t^{\kappa/(1+z)-n-1}}{1+t} dt.$$
 (4. 12)

We prove Proposition 4.2 only for the case of $\kappa = 2$, because one can prove the assertions similarly for $\kappa > 2$ and $1 < \kappa < 2$ making use of (4.8), (4.9), (4.10), (4.11) and (4.12). For the case of $\kappa = 2$, by (4.11) with n = 1, we see that

$$a(z) \equiv E_0[e^{-zT}] = 1 - 2z + b(z) + O(z^2), \quad z \downarrow 0,$$

and that

$$b(z) \sim 2z^2 \log \frac{1}{z}, \quad z \downarrow 0.$$

Note that we have by (4.9)

$$\psi(z) = -\frac{1}{1+2z}\frac{a(z)-1}{z} + \frac{2-a(z)}{1+2z},$$

and therefore $\psi(0+)=3$. Consequently, we have

$$\frac{\psi(z) - \psi(0+)}{z} = -\frac{1}{1+2z} \left(\frac{b(z)}{z^2} + O(1) \right),$$

$$\sim -2 \log \frac{1}{z}.$$

Hence, we conclude that

$$\frac{v(z) - v(0+)}{z} \sim -\frac{2}{v(0+)^2} \log \frac{1}{z} = -\frac{2}{9} \log \frac{1}{z}, \quad z \downarrow 0.$$

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