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Markov process representation of semigroups whose generators include negative rates

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Abstract

Generators of Markov processes on a countable state space can be represented as finite or infinite matrices. One key property is that the off-diagonal entries corresponding to jump rates of the Markov process are non-negative. Here we present stochastic characterizations of the semigroup generated by a generator with possibly negative rates. This is done by considering a larger state space with one or more particles and antiparticles, with antiparticles being particles carrying a negative sign.

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1 Introduction

Consider the generator L of a Markov jump process $(X_t)_{t\geq 0}$ on a countable state space E. It is characterized by jump rates r(x,y) for jumps from x to $y,x\neq y$, r(x,x)=0, and for $f:E\to\mathbb{R}$

$$Lf(x) = \sum_{y \in E} r(x, y)[f(y) - f(x)]. \tag{1.1}$$

The relationship between the probabilistic process $(X_t)_{t\geq 0}$, its semi-group $(P_t)_{t\geq 0}$ with $P_t f(x) = \mathbb{E}_x f(X_t)$ and generator describing the rules for jumps is very fruitful. One essential restriction is that the jump rates are non-negative. If r(x,y) < 0 is allowed, then (1.1) is still a perfectly valid operator which under reasonable conditions will be the generator of a semi-group $S_t = e^{tL}$, but the probabilistic interpretation is lost. The aim of this note is to recover some probabilistic meaning.

Before we go into the details let us remind us of some basic facts. In the probabilistic setting, the generator L is usually characterized via its jump rates r(x,y). If we consider L as matrix, then its off-diagonal entries are given by r(x,y), while the diagonal is given by $-\sum_{y:x\neq y} r(x,y)$. The fact that a Markov generator as a matrix has zero sum rows stems form the preservation of mass. The off diagonal entry r(x,y) is the parameter of the exponential waiting time for a jump from x to y. When presented with a matrix A where the diagonal entries do not match $-\sum_{y:x\neq y} r(x,y)$ but the off-diagonal entries are non-negative, then the deviation can be split off into a potential V, writing

$$Af(x) = (L+V)f(x) = \sum_{y \in E} r(x,y)[f(y) - f(x)] + V(x)f(x)$$
 (1.2)

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with V a diagonal matrix and L a Markov generator. The potential term V(x) has the probabilistic interpretation of a branching or killing rate, and the corresponding semigroup has (assuming for simplicity finite row sum norm of A) the explicit probabilistic form

$$e^{tA}f(x) = \mathbb{E}_x f(X_t) e^{\int_0^t V(X_u) du}$$
(1.3)

with X_t the Markov process generated by L. Equation (1.3) is sometimes referred to as the Feynman-Kac formula. The basic intuition behind this formula is that it represents the expectation of particles moving independently according to L, and which are killed or branch into two at rate |V(x)|, with negative V implying killing.

We will build on this intuition to deal with negative jump rates, which should represent both movement and killing. We can consider a regular jump event from x to y via positive rates r(x,y)>0 as the killing of a particle at x and creation of a particle at y. Correspondingly we will see that a 'jump' event from negative rates r(x,y)<0 is in some sense the opposite, the destruction of a particle at y and the creation of one at x. This runs into the problem that there might be no particle at y to destroy. We solve this by introducing anti-particles, and consider killing a particle at y the same as creating an anti-particle at y. In the following sections we will look at the details, with Theorem 2.1 corresponding to a single (anti-)particle like in (1.3) and Theorems 3.1 and Theorem 4.1 giving multi-particle formulations. Section 5 looks at an application to duality of Markov processes and Section 6 gives the simple example of a double Laplacian.

2 Switching between particles and antiparticles

Let us write $r^+(x,y) = \max(r(x,y),0)$ and $r^-(x,y) = \max(-r(x,y),0)$, and consider the Markov process $(\widehat{X}_t,Z_t)_{t\geq 0}$ on $E\times\{-1,+1\}$ with generator

$$\widehat{L}f(x,s) = \sum_{y \in E} r^{+}(x,y) \left[f(y,s) - f(x,s) \right] + \sum_{y \in E} r^{-}(x,y) \left[f(y,-s) - f(x,s) \right]. \tag{2.1}$$

We interpret \widehat{X}_t as the position of the Markov process, and Z_t indicates whether it is a particle $(Z_t=+1)$ or an anti-particle $(Z_t=-1)$. Then the first sum describes just the regular change of position via jumps utilizing the rates r^+ . The second sum similarly describes movement, but whenever the particle jumps according to the rates r^- , the state also changes from particle to anti-particle or vice versa. We can now present a stochastic representation of the semi-group generated by an arbitrary matrix with finite supremum norm.

Theorem 2.1. Let A be of the form

$$Af(X) = \sum_{y \in E} r(x, y)[f(y) - f(x)] + V(x)f(x)$$

and assume $\sup_{x\in E}\sum_{y\in E}|r(x,y)|<\infty$, $\sup_{x\in E}|V(x)|<\infty$, r(x,x)=0. Then A is a bounded operator w.r.t. the supremum-norm, $S_t=e^{tA}$ is well-defined and for any $f:E\to\mathbb{R}$ bounded, we have

$$S_t f(x) = \mathbb{E}_{x,+1} \left[Z_t f(\hat{X}_t) e^{2 \int_0^t \sum_{y \in E} r^-(\hat{X}_u, y) + V(\hat{X}_u) du} \right]. \tag{2.2}$$

Proof. Write $\hat{f}(x,s)=sf(x)$ and $\hat{V}(x,s)=2\sum_{y\in E}r^-(x,y)+V(x)$. Then the right hand side of (2.2) is the Feynman-Kac formulation of the solution of

$$\begin{cases} \frac{\partial \phi_t}{\partial t}(x,s) = \widehat{L}\phi_t(x,s) + \widehat{V}(x,s)\phi_t(x,s), \\ \phi_0 = \widehat{f}. \end{cases}$$
 (2.3)

On the other hand, $\tilde{\phi}_t(x,s) := sS_tf(x)$ also satisfies

$$\frac{\partial \tilde{\phi}_t}{\partial t}(x,s) = sAS_t f(x) = \hat{L}\tilde{\phi}_t(x,s) + \hat{V}(x,s)\tilde{\phi}_t(x,s),$$

and since $\tilde{\phi}_0 = \phi_0$ the claim (2.2) follows.

3 Branching particles and antiparticles

Consider a system of particles $\eta_t^+ \in \mathbb{N}_0^E$ and antiparticles $\eta_t^- \in \mathbb{N}_0^E$, where $\eta_t^\pm(x)$ is the number of particles/anti-particles at site x and time t. These particles move independently with jump rates $r^+(x,y)$. Additionally there is the following branching mechanism: a particle at site x branches into two particles at x and one anti-particle at site y at rate $r^-(x,y)$. The same is true for antiparticles at x, which branch into two at x plus a particle at y. The generator describing the movement and branching of particles is

$$L_{+}^{\uparrow}f(\eta^{+},\eta^{-}) = \sum_{x,y} r^{+}(x,y)\eta^{+}(x)[f(\eta^{+} + \delta_{y} - \delta_{x},\eta^{-}) - f(\eta^{+},\eta^{-})]$$
 (3.1)

$$+\sum_{x,y}r^{-}(x,y)\eta^{+}(x)[f(\eta^{+}+\delta_{x},\eta^{-}+\delta_{y})-f(\eta^{+},\eta^{-})]. \tag{3.2}$$

The first line of the generator describes the movement of particles. The rate $r^+(x,y)\eta^+(x)$ is the total rate that one of the $\eta^+(x)$ many particles at x jumps from x to y. After this jump there is one less particle at x and one more at y, making the new particle configuration $\eta^+ + \delta_y - \delta_x$. The configuration of anti-particles η^- is unchanged. The second line describes the branching mechanism, with $r^-(x,y)\eta^+(x)$ the aggregate rate that one of the particles at x turns into two particles at x and one anti-particle at y. In total the result is one more particle at x and one more anti-particle at y, resulting in the change $(\eta^+, \eta^-) \to (\eta^+ + \delta_x, \eta^- + \delta_y)$.

The generator describing the movement and branching of anti-particles is analogous, with the roles of particles and anti-particles reversed:

$$L_{-}^{\uparrow}f(\eta^{+},\eta^{-}) = \sum_{x,y} r^{+}(x,y)\eta^{-}(x)[f(\eta^{+},\eta^{-} + \delta_{y} - \delta_{x}) - f(\eta^{+},\eta^{-})]$$
 (3.3)

$$+\sum_{x,y}r^{-}(x,y)\eta^{-}(x)[f(\eta^{+}+\delta_{y},\eta^{-}+\delta_{x})-f(\eta^{+},\eta^{-})]. \tag{3.4}$$

The generator $L^\uparrow=L_+^\uparrow+L_-^\uparrow$ then describes the total system. This system is well-defined under the assumption that $\sup_{x\in E}\sum_{y\in E}|r(x,y)|=M<\infty$, which guarantees that there is no explosion: if $N_t=\sum_x\eta_t^+(x)+\sum_x\eta_t^-(x)$ is the total number of particles and antiparticles in the system, then N_t is dominated by a jump process with jumps from n to n+2 at rate nM, which leads to exponential growth but no explosion. Also note that under the dynamics the number $\sum_x\eta_t^+(x)-\sum_x\eta_t^-(x)$ is preserved in time. In particular, for the system starting with a single particle at x, i.e., $\eta_0^+=\delta_x$ and $\eta_0^-=0$, the sum is always 1.

Theorem 3.1. Assume $\sup_{x \in E} \sum_{y \in E} |r(x,y)| < \infty$. Given $f: E \to \mathbb{R}$ bounded, define

$$f^{\uparrow}(\eta^+, \eta^-) = \sum_{x \in E} (\eta^+(x) - \eta^-(x)) f(x). \tag{3.5}$$

Then the semigroup S_t generated by (1.1) has the stochastic description

$$S_t f(x) = \mathbb{E}_{(\delta_x, 0)} f^{\uparrow}(\eta_t^+, \eta_t^-). \tag{3.6}$$

Proof. Let $(\eta_t^{+,i},\eta_t^{-,i})_{t\geq 0},\ i=1,...,n$ be independent realizations of the particle system started at $(\eta_0^{+,i},\eta_0^{-,i})$. Then, by the independence of the branching and movement of particles, $(\sum_{i=1}^n \eta_t^{+,i}, \sum_{i=1}^n \eta_t^{-,i})_{t\geq 0}$ has the same law as a system started in $(\sum_{i=1}^n \eta_0^{+,i}, \sum_{i=1}^n \eta_0^{-,i})$. As a consequence, since f^\uparrow is linear in η^+,η^- , and anti-symmetric under exchange of η^+ and η^- ,

$$\mathbb{E}_{\eta_0^+, \eta_0^-} f^{\uparrow}(\eta_t^+, \eta_t^-) = \sum_x \eta_0^+(x) \mathbb{E}_{\delta_x, 0} f^{\uparrow}(\eta_t^+, \eta_t^-) + \sum_x \eta_0^-(x) \mathbb{E}_{0, \delta_x} f^{\uparrow}(\eta_t^+, \eta_t^-)$$
(3.7)

$$= \sum_{x} \eta_0^+(x) \mathbb{E}_{\delta_x,0} f^{\uparrow}(\eta_t^+, \eta_t^-) - \sum_{x} \eta_0^-(x) \mathbb{E}_{\delta_x,0} f^{\uparrow}(\eta_t^+, \eta_t^-)$$
 (3.8)

and in particular

$$\mathbb{E}_{2\delta_x,\delta_y} f^{\uparrow}(\eta_t^+,\eta_t^-) - \mathbb{E}_{\delta_x,0} f^{\uparrow}(\eta_t^+,\eta_t^-) = \mathbb{E}_{\delta_x,0} f^{\uparrow}(\eta_t^+,\eta_t^-) - \mathbb{E}_{\delta_y,0} f^{\uparrow}(\eta_t^+,\eta_t^-). \tag{3.9}$$

If we write $u_t(x) = \mathbb{E}_{\delta_x,0} f^{\uparrow}(\eta_t^+, \eta_t^-)$, then

$$\frac{d}{dt}u_t(x) = \left[L^{\uparrow}\mathbb{E}.f^{\uparrow}(\eta_t^+, \eta_t^-)\right](\delta_x, 0) \tag{3.10}$$

$$= \sum_{y} r^{+}(x, y) \left[\mathbb{E}_{\delta_{y}, 0} f^{\uparrow}(\eta_{t}^{+}, \eta_{t}^{-}) - \mathbb{E}_{\delta_{x}, 0} f^{\uparrow}(\eta_{t}^{+}, \eta_{t}^{-}) \right]$$
(3.11)

$$+\sum_{y}r^{-}(x,y)\left[\mathbb{E}_{2\delta_{x},\delta_{y}}f^{\uparrow}(\eta_{t}^{+},\eta_{t}^{-})-\mathbb{E}_{\delta_{x},0}f^{\uparrow}(\eta_{t}^{+},\eta_{t}^{-})\right]$$
(3.12)

$$= \sum_{y} r^{+}(x,y) \left[\mathbb{E}_{\delta_{y},0} f^{\uparrow}(\eta_{t}^{+}, \eta_{t}^{-}) - \mathbb{E}_{\delta_{x},0} f^{\uparrow}(\eta_{t}^{+}, \eta_{t}^{-}) \right]$$
(3.13)

$$+\sum_{y}r^{-}(x,y)\left[\mathbb{E}_{\delta_{x},0}f^{\uparrow}(\eta_{t}^{+},\eta_{t}^{-})-\mathbb{E}_{\delta_{y},0}f^{\uparrow}(\eta_{t}^{+},\eta_{t}^{-})\right]$$
(3.14)

$$= Lu_t(x). (3.15)$$

Hence $u_t(x)$ is the unique solution of

$$\begin{cases} \frac{\partial u_t}{\partial t}(x) = Lu_t(x), \\ u_0 = f(x). \end{cases}$$
 (3.16)

Remark 3.2. Theorem 3.1 assumes for simplicity and readability that there is no potential. The presence of a potential V like in Theorem 2.1 would mean that there is in addition branching and annihilation of particles and antiparticles via

$$L_V^{\uparrow} f(\eta^+, \eta^-) = \sum_x V^+(x) \eta^+(x) [f(\eta^+ + \delta_x, \eta^-) - f(\eta^+, \eta^-)]$$
 (3.17)

$$+\sum_{x}V^{+}(x)\eta^{-}(x)[f(\eta^{+},\eta^{-}+\delta_{x})-f(\eta^{+},\eta^{-})]$$
 (3.18)

$$+\sum V^{-}(x)\eta^{+}(x)[f(\eta^{+}-\delta_{x},\eta^{-})-f(\eta^{+},\eta^{-})]$$
 (3.19)

$$+\sum_{x}V^{-}(x)\eta^{-}(x)[f(\eta^{+},\eta^{-}-\delta_{x})-f(\eta^{+},\eta^{-})]$$
 (3.20)

meaning both particles and anti-particles individually branch into two at rate V^+ or are killed at rate V^- . It can be easily verified that $[L_V^{\uparrow}\mathbb{E}.f^{\uparrow}(\eta_t^+,\eta_t^-)](\delta_x,0)=V(x)u_t(x)$.

4 Branching and annihilating particles and antiparticles

The process in Section 3 tends to have an exponentially growing number of particles. It turns out that we can introduce annihilation of particles and antiparticles to reduce this number. We do so by letting any pair of particle and antiparticle which are at the same site annihilate at rate $\lambda \in [0,\infty]$, where infinite rate corresponds to instant annihilation. Let

$$L^{\uparrow,\lambda}f(\eta^+,\eta^-) = L^{\uparrow}f(\eta^+,\eta^-) + \lambda \sum_x \eta^+(x)\eta^-(x) \left[f(\eta^+ - \delta_x,\eta^- - \delta_x) - f(\eta^+,\eta^-) \right] \ \ \textbf{(4.1)}$$

be the generator of the particle system which includes annihilation.

Theorem 4.1. Theorem 3.1 is also valid when there is annihilation for any $\lambda \in (0, \infty]$.

Proof. Write $P_t^{\uparrow,\lambda}f(\eta^+,\eta^-)=\mathbb{E}_{\eta^+,\eta^-}f(\eta_t^+,\eta_t^-)$ for the semigroup generated by $L^{\uparrow,\lambda}$, with $\lambda=0$ being the system without annihilation. By (3.8), if $\eta^+(x)>0$ and $\eta^-(x)>0$,

$$P_t^{\uparrow,0} f^{\uparrow}(\eta^+, \eta^-) = P_t^{\uparrow,0} f^{\uparrow}(\eta^+ - \delta_x, \eta^- - \delta_x).$$

Hence

$$(L^{\uparrow,\lambda} - L^{\uparrow,0})P_t^{\uparrow,0}f^{\uparrow}(\eta^+, \eta^-) = 0$$
(4.2)

and it follows that $P_t^{\uparrow,\lambda} f^{\uparrow} = P_t^{\uparrow,0} f^{\uparrow}$.

5 Applications to duality of Markov processes

A very brief introduction to duality of Markov processes is as follows. Two Markov processes $(X_t)_{t\geq 0}$ and $(Y_t)_{t\geq 0}$ on state spaces E and F are said to be dual with duality function $H: E\times F\to \mathbb{R}$, if for all $x\in E$ and $y\in F$,

$$\mathbb{E}_x H(X_t; y) = \mathbb{E}_y H(x; Y_t). \tag{5.1}$$

A sufficient condition is that the generators L_X and L_Y satisfy

$$[L_X H(\cdot; y)](x) = [L_Y H(x; \cdot)](y), \quad \forall \ x \in E, y \in F.$$

$$(5.2)$$

Duality has proven fruitful in many applications. For a survey on duality, see [2]. The challenge with duality is that given a Markov process X_t of interest, how to find a Markov process Y_t and duality function H so that (5.1) holds. One can make an educated guess on H, and then find a generator L_Y which satisfies (5.2). Or one can use symmetries of L_X to identify a suitable Lie algebra representation whose building blocks can build L_X , and then find a dual representation, which then allows to build L_Y , see [1] and [3] for an introduction to this method. However, neither method guarantees that the dual generator L_Y is actually a Markov generator. If F is countable, as is the case in many applications of duality, then L_Y can be represented as a finite or infinite matrix. A stochastic representation of the semigroup generated by such an L_Y is desirable, and with Theorem 2.1, Theorem 3.1 or Theorem 4.1 this is possible.

Theorem 5.1. Assume that there is a duality function H and generator L_Y satisfying (5.2), with F countable. Further assume that the matrix representation of L_Y has row sums 0, so that it can be written in the form of (1.1), and $\sup_{y\in F}\sum_{z\in F}|r(y,z)|<\infty$. Then the Markov process $(X_t)_{t>0}$ is dual to the process $(\eta_t^+,\eta_t^-)_{t>0}$ with duality function

$$H^{\uparrow}(x; \eta^+, \eta^-) = \sum_{y \in F} (\eta^+(y) - \eta^-(y)) H(x; y).$$

Here $(\eta_t^+, \eta_t^-)_{t\geq 0}$ is the branching (and annihilating) particle system introduced in sections 3 and 4, with arbitrary annihilation rate $\lambda \in [0, \infty]$. In other words

$$\mathbb{E}_x H^{\uparrow}(X_t; (\eta^+, \eta^-)) = \mathbb{E}_{\eta^+, \eta^-} H^{\uparrow}(x; (\eta_t^+, \eta_t^-)). \tag{5.3}$$

Proof. By the proof of Theorem 4.1 the right hand side of (5.3) does not depend on the annihilation rate, so we can restrict ourself to the case of no annihilation. By (5.2) we have $\mathbb{E}_x H(X_t;y) = [S_t H(x;\cdot)](y)$, where S_t is the semigroup generated by L_Y . Then, by Theorem 3.1, we have

$$\mathbb{E}_x H^{\uparrow}(X_t; (\delta_y, 0)) = \mathbb{E}_x H(X_t; y) = [S_t H(x; \cdot)](y) = \mathbb{E}_{\delta_y, 0} H^{\uparrow}(x; (\eta_t^+, \eta_t^-)). \tag{5.4}$$

Finally, with (3.8) we can extend the above from $(\delta_y, 0)$ to arbitrary starting configurations.

6 Example: double Laplacian on the integers

Let $\Delta f(x) = \frac{1}{2}f(x+1) - f(x) + \frac{1}{2}f(x-1)$ be the discrete Laplacian on \mathbb{Z} . Then the double Laplacian is given by

$$\Delta \Delta f(x) = \frac{1}{4} (f(x+2) - f(x)) + \frac{1}{4} (f(x-2) - f(x))$$
 (6.1)

$$-(f(x+1) - f(x)) - (f(x-1) - f(x)), \tag{6.2}$$

which is of the form (1.1) with negative rates. Let S_t be the semigroup generated by the double Laplacian $\Delta\Delta$. We will apply Theorem 2.1. So let \widehat{X} be the random walk on \mathbb{Z} which performs the jumps ± 1 at rate 1 and ± 2 at rate $\frac{1}{4}$. Since jumps using the rates r^- involve flipping the sign of Z_t , we have that $Z_t = (-1)^{N_t}$, where N_t is the number of nearest neighbour jumps performed by \widehat{X}_t . Note that N_t is even iff $\widehat{X}_t - \widehat{X}_0$ is even. Hence

$$Z_t = 2\mathbb{1}_{N_t \text{ is even}} - 1 = 2\mathbb{1}_{\widehat{X}_t - \widehat{X}_0 \text{ is even}} - 1.$$
 (6.3)

Finally we observe that by spatial homogeneity $\sum_{y} r^{-}(x,y) = 2$. By Theorem 2.1,

$$S_t f(x) = e^{4t} \mathbb{E}_x \left(Z_t f(\widehat{X}_t) \right). \tag{6.4}$$

Note that N_t is Poisson (2t)-distributed, and therefore $\mathbb{P}(N_t$ is even) = $\frac{1}{2}(1+e^{-4t})$ and $\mathbb{E}Z_t=e^{-4t}$. Alternatively, $\mathbb{E}Z_t=e^{-4t}$ follows from (6.4) applied to the constant function 1, since $S_t\mathbf{1}=\mathbf{1}$. For a more complex example consider f of the form $f(x)=g(x)\mathbb{1}_{x \text{ is even}}$. Then, by (6.3) and (6.4),

$$S_t f(x) = \begin{cases} \frac{1}{2} (e^{4t} + 1) \mathbb{E}_x [g(\widehat{X}_t) | \widehat{X}_t \text{ even}], & x \text{ even}; \\ -\frac{1}{2} (e^{4t} - 1) \mathbb{E}_x [g(\widehat{X}_t) | \widehat{X}_t \text{ odd}], & x \text{ odd.} \end{cases}$$
(6.5)

The conditional expectations are reasonably well approximated by integrating g against a normal distribution with variance $\mathrm{Var}(\widehat{X}_t)=4t$ assuming g is smooth enough and t not too small.

7 Example of all rates negative

Consider the operator of the form (1.2) with all $r(x,y) \leq 0$. We make the simplifying assumption that there are constants λ_1,λ_2 so that $\sum_y r^-(x,y) = \lambda_1$ and $V(x) = \lambda_2$ for all

x. Then, by Theorem 2.1,

$$e^{-(2\lambda_1 + \lambda_2)t} e^{At} f(x) = \mathbb{E}_{x,+1} \left(Z_t f(\widehat{X}_t) \right)$$

$$(7.1)$$

$$= \mathbb{E}_{x,+1} \left[f(\widehat{X}_t) \mid N_t \text{ even} \right] \mathbb{P}(N_t \text{ even}) - \mathbb{E}_{x,+1} \left[f(\widehat{X}_t) \mid N_t \text{ odd} \right] \mathbb{P}(N_t \text{ odd}), \tag{7.2}$$

where N_t counts the number of jumps of \widehat{X}_t . Since $\sum_y r^-(x,y) = \lambda_1$ it follows that N_t is Poisson $(\lambda_1 t)$ -distributed and $\mathbb{P}(N_t \text{ even}) = \frac{1}{2}(1 + e^{2\lambda_1 t})$.

If we assume that \hat{X}_t has a stationary distribution μ then it is reasonable to write

$$\mathbb{E}_{x,+1}\left[f(\widehat{X}_t) \mid N_t \text{ even}\right] = \mu(f) + b_t^e(x); \tag{7.3}$$

$$\mathbb{E}_{x,+1}\left[f(\widehat{X}_t) \mid N_t \text{ odd}\right] = \mu(f) + b_t^o(x). \tag{7.4}$$

If \widehat{X}_t is converging exponentially fast to μ the error terms $b_t^e(x)$ and $b_t^o(x)$ will be decaying exponentially at some rate $0 \le \nu \le \lambda_1$ (if the Markov process is on a bipartite graph ν can be 0 even if convergence to μ is exponentially fast, and the rate is no larger than λ_1 since that the exit rate for a single site). Then, continuing from (7.1),

$$e^{At}f(x) = e^{(2\lambda_1 + \lambda_2)t} \frac{b_t^e(x) - b_t^o(x)}{2} + e^{\lambda_2 t} \left(\mu(f) + \frac{b_t^e(x) + b_t^o(x)}{2} \right). \tag{7.5}$$

Therefore typically the first term is the dominant term, and $|e^{At}f|$ grows at rate at most $2\lambda_1 + \lambda_2 - \nu$, which is slower than what (7.1) or the supremum norm $||A||_{\infty} = |\lambda_1 + \lambda_2| + \lambda_1$ suggest. More details depend on a more sophisticated analysis using specifics of f and \hat{X}_t , for example by finding cancellations in $b_t^e(x) - b_t^o(x)$.

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