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J. Math. Comput. Sci. 3 (2013), No. 2, 569-576

ISSN: 1927-5307

THE STUDY OF RAY-KNIGHT COMPACTIFICATION ON TRANSFER FUNCTION

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Abstract: In order to construct a Markov chain of strong Markov, we need the state space of the compactification. The paper uses the properties of the resolvent operator to study some properties of Ray-Knight compactifications from a given transfer function on state space E .

Key words: transfer function; Ray resolvent; Ray-Knight Compactifications

2010 Mathematical Subject Classification: 60J10

1 Introduction

Construction of the transfer function of the given E is an important part of research of Markov chain. The state space in the canonical chain on $E \cup \infty$ is very simple, but its orbit has only lower semicontinuity, which just keep part of the strong Markov. In order to construct a Markov chain of strong Markov, the State space of the compactification is needed. We will study the properties of Ray-Knight Compactifications from a given transfer function on state space E .

2 Preliminary knowledge

Let $E = \{1, 2, \dots\}$, and the topology on E is discrete topology, then E is

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Received February 8, 2013

L.C.C.B(Locally compact and has a countable topological group). The elements of E are called state. \mathcal{E} is a Borel algebra on E , (E, \mathcal{E}) is a topological space, and $C_b(E)$ represents all bounded continuous functions. If E is a compact metric space, $C_b(E)$ abbreviated $C(E)$. Remove the dense subset $\{g_m\}_{m=1}^{\infty}$ of \mathfrak{R} , set $d(x, y) = \sum_{m=1}^{\infty} \frac{1}{2^m} \Lambda |g_m(x) - g_m(y)|, \forall x, y \in E$, $d(\cdot, \cdot)$ is the metric on E , completion of E under $d(\cdot, \cdot)$ written \bar{E} . It is obvious that \bar{E} is a compact metric space. The metric on \bar{E} still denoted as $d(\cdot, \cdot)$. $(U^\alpha)_{\alpha>0}$ is the Ray resolvent operator on \bar{E} , D is a no branch point set.

Definition 1: (\bar{E}, d) is called Ray-Knight compactification on E .

Definition2: If $p_{ij}(t) \geq 0; \sum_{k=1}^{\infty} p_{ik}(t) \leq 1; p_{ij}(t+s) = \sum_{k=1}^{\infty} p_{ik}(t)p_{kj}(s)$

establish, then the function $P(t) = (p_{ij}(t))_{i,j \in E}$ on $[0, \infty)$ is called the transfer function on E .

If $\lim_{t \rightarrow 0} p_{ij}(t) = p_{ij}(0) = \delta_{ij}$, then $P(t)(t \geq 0)$ is called the standard. If

$\forall i \in E, \sum_{k=1}^{\infty} p_{ik}(t) = 1$, Then $P(t)(t \geq 0)$ is said to be honest.

Definition 3: If $0 \leq q_{ij} < \infty, 0 \leq q_i \equiv -q_{ii} \leq \infty, \sum_{k \neq i} q_{ik} \leq q_i, \forall i, j \in E$, then

$Q = (q_{ij})_{i,j \in E}$ is called density matrix of $P(t)(t \geq 0)$. If $q_i < \infty$, then i is called stable state of $P(t)$. Otherwise i is called instantaneous state of $P(t)$. If $\sum_{k \neq i} q_{ik} = q_i$, then i is called conservative state of $P(t)$. If all States are stable, then $P(t)$ is said to be fully stabilized. If all States are conservative, then $P(t)$ is

said to be fully conservative.

Definition 4: Let $P(t)(t \geq 0)$ is the transfer function on E , $R_{ij}(\lambda)$ is called resolvent of $P(t)(t \geq 0)$. Among then $R_{ij}(\lambda) = \int_0^\infty e^{-\lambda t} p_{ij}(t) dt, i, j \in E, \lambda > 0$.

Lemma 1: $R_{ij}(\lambda)$ is the resolvent of transfer function $P(t)(t \geq 0)$, if and only if $\lambda \sum_{k \in E} R_{ik}(\lambda) \leq 1; R_{ij}(\lambda) - R_{ij}(\mu) + (\lambda - \mu) \sum_{k \in E} R_{ik}(\lambda) R_{kj}(\mu) = 0$;

$$\lim_{\lambda \rightarrow \infty} \lambda R_{ij}(\lambda) = \delta_{ij}; \lim_{\lambda \rightarrow \infty} \lambda [\lambda R_{ij}(\lambda) - \delta_{ij}] = q_{ij}.$$

Proof: Reference[2]

Lemma 2 If $Q = (q_{ij})_{i,j \in E}$ is density matrix of $P(t)(t \geq 0)$, and $P(t)$ is fully stabilized, then

$$p'_{ij} \geq -q_i p_{ij}(t) + \sum_{k \neq i} q_{ik} p_{kj}(t)$$

$$p'_{ij} \geq -p_{ij}(t) q_j + \sum_{k \neq j} p_{ik}(t) q_{kj}$$

Proof: From the definition of density matrix and the whole stability can be directly obtained the conclusion.

Note1 If the two formulas in lemma 2 an equality, we get two group of linear differential equation group, Are called backward equations and the forward equations.

Lemma 3 Let $Q = (q_{ij})_{i,j \in E}$ is the whole stability density matrix of $P(t)(t \geq 0)$, set

$$f_{ij}^{(0)}(t) = \delta_{ij} e^{-q_j t}, f_{ij}^{(n)}(t) = \sum_{k \neq i} \int_0^t e^{-q_i s} q_{ik} f_{kj}^{(n-1)}(t-s) ds, n = 1, 2, \dots$$

$$f_{ij}^{(0)}(t) = \delta_{ij} e^{-q_i t}, f_{ij}^{(n)}(t) = \sum_{k \neq j} \int_0^t f_{ik}^{(n-1)}(s) q_{kj} e^{-q_j(t-s)} ds, n = 1, 2, \dots$$

$$p_{ij}^{\min}(t) = \sum_{n=0}^{\infty} f_{ij}^{(n)}(t)$$

Then (1) $P^{\min}(t) = (p_{ij}^{\min}(t))_{i,j \in E}$ is transfer function on E, its density matrix is Q.

(2) $P^{\min}(t)$ is minimum, that is $p_{ij}(t) \geq p_{ij}^{\min}(t), \forall i, j \in E, t \geq 0$.

(3) $P^{\min}(t)$ satisfy the forward equations and backward equations.

Proof:Reference[1]

Note2: $P^{\min}(t)$ is called the minimum transfer function.

3 Main results

If the transfer function $P(t)$ is not honest, set $\Delta \notin E$, and

$$E_{\Delta} = E \cup \{\Delta\}, p_{\Delta\Delta}(t) = 1, p_{\Delta i}(t) = 0, p_{i\Delta} = 1 - \sum_{k \in E} p_{ik}(t), \forall i \in E$$

Then $P(t) = (p_{ij}(t))_{i,j \in E_{\Delta}}$ is the honest transfer function on E_{Δ} , so the $P(t)$ can be transformed into $P(t)$ to discuss.

Let $P(t) = (p_{ij}(t))_{i,j \in E}$ is the honest transfer function on E. $\forall f \in M$, the function $i \mapsto \sum_{k \in E} R_{ik}(\lambda) f(k)$ on E is noted $R_{\lambda} f$.

Theorem 1: $(R_{\lambda})_{\lambda > 0}$ is the Markov resolvent on E, and have:

(1) E is L.C.C.B topological space.

(2) $\{R_{\lambda}\}_{\lambda > 0}$ is Markov, and $R_{\lambda} C_b(E) \subset C_b(E), \forall \lambda > 0$.

(3) For any $f \in C_b(E)$, and $x \in E, \lim_{\lambda \rightarrow \infty} \lambda R_{\lambda} f(x) = f(x)$

Proof: (1) and (2) is obvious. we proof (3):

$$\begin{aligned} \lambda R_{\lambda} f(i) &= \lambda \sum_{k \in E} R_{ik}(\lambda) f(k) = \lambda \int_0^{\infty} e^{-\lambda t} \sum_{k \in E} p_{ik}(t) f(k) dt \\ &= \lambda \int_0^{\infty} e^{-\lambda t} p_{ii}(t) f(i) dt + \lambda \int_0^{\infty} e^{-\lambda t} \sum_{k \neq i} p_{ik}(t) f(k) dt \end{aligned}$$

Because $\lim_{t \rightarrow 0} p_{ii}(t) = 1$, when $\lambda \rightarrow \infty$, $\lambda \int_0^{\infty} e^{-\lambda t} p_{ii}(t) f(i) dt \rightarrow f(i)$;

$$\lambda \int_0^{\infty} e^{-\lambda t} \sum_{k \neq i} p_{ik}(t) f(k) dt \leq \|f\| \lambda \int_0^{\infty} e^{-\lambda t} (1 - p_{ii}(t)) dt \rightarrow 0$$

So $\lim_{\lambda \rightarrow \infty} \lambda R_\lambda f(i) = f(i)$.

Theorem 2: Let $(\bar{E}, d(\cdot, \cdot))$ is Ray-Knight compactifications on E , then:

(1) $E \subseteq D$

(2) For any $i \in E, \alpha > 0, U^\alpha(i, \bar{E} \setminus E) = 0$, and for any $k \in E$,

$$U^\alpha(i, \{k\}) = R_{ik}(\alpha)$$

(3) For any $i \in E, t > 0, P_t(i, \bar{E} \setminus E) = 0$, and for any $k \in E$,

$$P_t(i, \{k\}) = p_{ik}(t)$$

Proof : (1) For any $i \in E, \alpha > 0, f, g \in \mathfrak{R}$, because $U^\alpha f, U^\alpha g$ are $R_\alpha f, R_\alpha g$ expansion in the E , so

$$U^\alpha(f - g)(i) = R_\alpha f(i) - R_\alpha g(i) = \sum_{k \in E} R_{ik}(\alpha)[f(k) - g(k)],$$

For any $f \in C(\bar{E})$, $U^\alpha f(i) = \sum_{k \in E} R_{ik}(\alpha)f(k) = R_\alpha f(i)$, because R_α meet three conclusions of the theorem 1. so we have $\lim_{\alpha \rightarrow \infty} \alpha U^\alpha f(i) = f(i)$, so that

$i \in D$. For i is optional nature, we have $E \subseteq D$.

(2) For bounded measurable function f on any \bar{E} , it is obvious:

$$U^\alpha f(i) = R_\alpha f(i) = \sum_{k \in E} R_{ik}(\alpha)f(k)$$

For any $k \in E$, Use the characteristic function $I_k(\cdot)$ of $\{k\}$ instead of f , we get $U^\alpha(i, \{k\}) = R_{ik}(\alpha)$. Both ends of the sum of k , we get

$$U^\alpha(i, E) = \sum_{k \in E} U^\alpha(i, \{k\}) = \sum_{k \in E} R_{ik}(\alpha) = \frac{1}{\alpha} = U^\alpha(i, \bar{E})$$

So that $U^\alpha(i, \bar{E} \setminus E) = 0$.

(3) Beacuse $(U^\alpha)_{\alpha>0}$ is Ray resolvent on \bar{E} , so for any $i, k \in E, f \in C(\bar{E}), \alpha > 0$,

$$\int_0^\infty e^{-\alpha t} P_t f(i) dt = U^\alpha f(i) = \sum_{k \in E} R_{ik}(\alpha) f(k) = \int_0^\infty e^{-\alpha t} \sum_{k \in E} p_{ik}(t) f(k) dt$$

For any $t_0 > 0, t > 0, h > 0$,

$$\begin{aligned} & \left| \sum_{k \in E} p_{ik}(t_0 + t + h) f(k) - \sum_{k \in E} p_{ik}(t_0 + t) f(k) \right| \\ & \leq \|f\| \sum_{k \in E} |p_{ik}(t_0 + t + h) - p_{ik}(t_0 + t)| \\ & = \|f\| \left[\sum_{k \in E} \left| \sum_{m \in E} p_{im}(t_0) (\sum_{l \in E} p_{ml}(h) p_{lk}(t) - p_{mk}(t)) \right| \right] \\ & \leq \|f\| \left[\sum_{k \in E} \sum_{m \in E} p_{im}(t_0) |p_{mm}(h) - 1| p_{mk}(t) \right] + \|f\| \left[\sum_{k \in E} \sum_{m \in E} \sum_{l \in E} p_{im}(t_0) p_{ml}(h) p_{lk}(t) \right] \\ & = 2 \|f\| \left[\sum_{m \in E} p_{im}(t_0) [1 - p_{mm}(h)] \right] \end{aligned}$$

By the $p_{ij}(t)$ standard and control convergence theorem^[5] to:

$$\lim_{h \rightarrow 0} \left[\sum_{k \in E} p_{ik}(t_0 + t + h) f(k) - \sum_{k \in E} p_{ik}(t_0 + t) f(k) \right] = 0$$

That is $t \mapsto \sum_{k \in E} p_{ik}(t) f(k)$ is continuous function on $(0, \infty)$, For α is optional nature, we have : $\forall t > 0, \int_{\bar{E}} P_t(i, dy) f(y) = P_t f(i) = \sum_{k \in E} p_{ik}(t) f(k)$. By the monotone class theorem^[5], The formula for arbitrary bounded measurable function on \bar{E} is also established.

$$\forall k \in E, \text{ let } f(\cdot) = I_k(\cdot), \text{ substitution } U^\alpha f(i) = \sum_{k \in E} R_{ik}(\alpha) f(k) = R_\alpha f(i),$$

we get $P_t(i, \{k\}) = p_{ik}(t)$, On both sides of the k sum, then $\sum_{k \in E} P_t(i, \{k\}) = 1$, so

$$P_t(i, \bar{E} \setminus E) = 0. \quad \square$$

Note3: Let $E_R = \{x \mid x \in \bar{E}, U^1(x, E) = 1\}$, it is clear that E_R is Borel subset on \bar{E} .

Theorem3: Let $x \in E_R$, then $\forall t > 0, P_t(x, \bar{E}) = P_t(x, E)$, and $\forall s \geq 0, k \in E$,

$$P_{t+s}(x, \{k\}) = \sum_{m \in E} P_t(x, \{m\}) p_{mk}(s).$$

Proof: because $\forall t > 0, P_t(x, E) \leq P_t(x, \bar{E}) = 1$, so

$$U^1(x, E) = \int_0^\infty e^{-t} P_t(x, E) dt \leq \int_0^\infty e^{-t} P_t(x, \bar{E}) dt = 1$$

By $x \in E_R, 1 = \int_0^\infty e^{-t} P_t(x, E) dt = \int_0^\infty e^{-t} P_t(x, \bar{E}) dt$, so $P_t(x, E) = P_t(x, \bar{E}) = 1$, then

$\exists \{t_n\}_{n=1}^\infty$, so that when $t \rightarrow 0$ and $\forall t_n, P_{t_n}(x, E) = 1$. For any $t > 0$, Let $t_n < t$, by The

properties^[6] of semigroup of $(P_s)_{s \geq 0}: P_t(x, E) = \int_{\bar{E}} P_{t_n}(x, dy) P_{t-t_n}(y, E)$

$$= \int_E P_{t_n}(x, dy) P_{t-t_n}(y, E) \sum_{k \in E} P_{t-t_n}(y, \{k\}) p_{tk}(t-t_n) = P_t(x, E) = 1$$

$$= P_t(x, E) = 1$$

We get: $P_t(x, E) = P_t(x, \bar{E}), \forall t > 0, s \geq 0, k \in E:$

$$P_{t+s}(x, \{k\}) = \int_{\bar{E}} P_t(x, dy) P_s(y, \{k\}) = \int_E P_t(x, dy) P_s(y, \{k\}) = \sum_{m \in E} P_t(x, \{m\}) p_{mk}(s)$$

$\forall x \in E_R, k \in E, t > 0, P_t(x, \{k\})$ is denoted $p_{xk}(t)$.

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