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ASYMPTOTIC ARC-SINE LAWS FOR FINITE-DIMENSIONAL INTERACTING DIFFUSIONS

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Abstract

We consider finite-dimensional interacting diffusions which are defined by adding a linear drift term to independent one dimensional diffusions. For these processes we prove that the distribution of the occupation time at the first quadrant converges to a generalized arc-sine law.

1. Introduction

Let S be a finite set, and let $A = \{A_{ij}\}_{i \neq j \in S}$ be a matrix with non-negative elements. Let us consider the following stochastic differential equation (SDE):

$$(1.1) \quad dX_i(t) = \alpha(X_i(t)) dB_i(t) + \sum_{j \in S} A_{ij}(X_j(t) - X_i(t)) dt, \quad (i \in S),$$

where $\{B_i(t)\}_{i \in S}$ is an independent system of one-dimensional standard Brownian motions.

Assume that $\alpha: \mathbb{R} \rightarrow \mathbb{R}_+$ is a Borel measurable function satisfying the following conditions:

[A-1] For some positive constant $C > 0$,

$$(1.2) \quad \alpha(x) \leq C(1 + |x|) \quad \text{for } x \in \mathbb{R}.$$

[A-2] For each compact set K , there exists a positive constant c_K such that $\alpha(x) \geq c_K$ ($x \in K$),

one can see by standard arguments to use the Girsanov theorem that for any initial distribution on \mathbb{R}^S , the SDE (1.1) has a unique weak solution, which defines a diffusion process $(X(t), P_x)$ on \mathbb{R}^S . We call the diffusion process a *finite-dimensional interacting diffusion*.

In this paper we are concerned with limiting distribution as $t \rightarrow \infty$ of the occupation time of $X(t)$ at the first quadrant $\mathbb{R}_+^S = [0, \infty)^S$ of \mathbb{R}^S

$$(1.3) \quad \frac{1}{t} \int_0^t I_{\mathbb{R}_+^S}(X(s)) ds.$$

In non-interacting case where $A = \{A_{ij}\}$ is absent, each coordinate process is a diffusion process $(X(t), P_x)$ on \mathbb{R} governed by the following SDE:

$$(1.4) \quad dX(t) = \alpha(X(t)) dB(t).$$

For the one-dimensional diffusion process $(X(t), P_x)$ governed by (1.4) Watanabe [5] proved that the distribution of

$$\frac{1}{t} \int_0^t I_{\mathbb{R}_+}(X(s)) ds$$

converges to a non-degenerate distribution as $t \rightarrow \infty$ if and only if

$$m_+(x) = \int_0^x \alpha(u)^{-2} du, \quad m_-(x) = \int_{-x}^0 \alpha(u)^{-2} du \quad (x \geq 0)$$

satisfy the following condition; for some $0 < p < 1$

$$(1.5) \quad m_{\pm}(x) = x^{1/p-1} K_{\pm}(x)$$

with slowly varying functions $K_+(x)$ and $K_-(x)$ at $x = \infty$ and

$$(1.6) \quad \lim_{x \nearrow \infty} \frac{K_+(x)}{K_-(x)} = b \in (0, \infty).$$

Then it holds that

$$\frac{1}{t} \int_0^t I_{\mathbb{R}_+}(X(s)) ds \xrightarrow{(d)} Y_{p,q} \quad (t \rightarrow \infty),$$

where q is given by

$$q = \frac{b^p}{1 + b^p} \in (0, \infty),$$

and $\xrightarrow{(d)}$ denotes convergence in distribution and $Y_{p,q}$ is a $[0, 1]$ -valued random variable with the Stieltjes transform given by

$$E \left[\frac{1}{u + Y_{p,q}} \right] = \frac{q(u+1)^{p-1} + (1-q)u^{p-1}}{q(u+1)^p + (1-q)u^p}, \quad u > 0.$$

The family $Y_{p,q}$, $0 < p \leq 1$, $0 < q < 1$, was introduced by Lamperti [2], of which distribution is called a *generalized arc-sine law*. In particular, the distribution of $Y_{1/2,1/2}$ is the arc-sine law, of which density function is given by

$$\frac{1}{\pi \sqrt{x(1-x)}}.$$

For general $0 < p < 1$ and $0 < q < 1$, $Y_{p,q}$ has the density $f_{p,q}(x)$ on $[0, 1]$;

$$f_{p,q}(x) = \frac{\sin p\pi}{\pi} \frac{q(1-q)x^{p-1}(1-x)^{p-1}}{q^2(1-x)^{2p} + (1-q)^2x^{2p} + 2q(1-q)x^p(1-x)^p \cos p\pi}.$$

For the finite-dimensional interacting diffusion $(X(t), P_x)$ governed by (1.1) we investigate the limiting distribution of (1.3) under the following condition:

[B-1] $\alpha(x)$ is regularly varying both at $x \rightarrow \infty$ and $x \rightarrow -\infty$ with the common exponent $-\infty < \gamma < 1/2$, and

$$\lim_{x \rightarrow \infty} \frac{\alpha(-x)}{\alpha(x)} = c \in (0, \infty).$$

[B-2] An $S \times S$ -matrix $A = \{A_{ij}\}_{i,j \in S}$, of which diagonal element is defined by

$$A_{ii} = - \sum_{j \in S, j \neq i} A_{ij} \quad (i \in S),$$

is irreducible.

We note that by [B-2]

$$Q_t = \exp tA$$

defines a transition matrix of an irreducible Markov process on S , so that there exists a probability vector $m = \{m_i\}_{i \in S}$ with $m_i > 0$ such that for some $\delta > 0$

$$(1.7) \quad |Q_t(i, j) - m_j| \leq e^{-\delta t} \quad (i, j \in S).$$

The main result of this paper is the following.

Theorem 1.1. *Assume the conditions [B-1] and [B-2]. Then*

$$(1.8) \quad \frac{1}{t} \int_0^t \delta_{X(s)} ds \xrightarrow{(d)} Y_{p,q} \delta_{+\infty} + (1 - Y_{p,q}) \delta_{-\infty} \quad (t \rightarrow \infty),$$

where $+\infty = \{x_i \equiv +\infty\}$, $-\infty = \{x_i \equiv -\infty\}$, $\delta_{X(s)}$, $\delta_{+\infty}$ and $\delta_{-\infty}$ stand for the one point mass at $X(s)$, $+\infty$ and $-\infty$ respectively, and $\xrightarrow{(d)}$ denotes the weak convergence as $\mathcal{P}([-\infty, \infty]^S)$ -valued random variables, and here p, q are given by

$$p = \frac{1}{2(1-\gamma)}, \quad q = \frac{c^{2p}}{1+c^{2p}}.$$

From Theorem 1.1 it follows immediately that

Corollary 1.2. *Assume the same assumptions as in Theorem 1.1. Then*

$$(1.9) \quad \frac{1}{t} \int_0^t I_{\mathbb{R}^S}(X(s)) ds \xrightarrow{(d)} Y_{p,q} \quad (t \rightarrow \infty).$$

The result of Theorem 1.1 can be interpreted as follows. Since S is a finite set, the effect of the interaction $A = \{A_{ij}\}$ is so strong that all component processes diverge to ∞ or $-\infty$ as $t \rightarrow \infty$ simultaneously. Hence the phenomena would be quite similar to the one-dimensional case. Nevertheless the one-dimensional analysis as in Watanabe [5] cannot be applied, so, in the next section, we will investigate a scaling limit for the finite-dimensional interacting diffusion $(X(t), P_x)$ on \mathbb{R}^S .

2. A scaling limit of $X(t)$

By the condition [B-1] $\alpha(x)$ has the following form;

$$\alpha(x) = |x|^\gamma L(x) \quad (|x| > 0),$$

where $L(x)$ is a slowly varying function both at ∞ and $-\infty$ satisfying that

$$\lim_{x \rightarrow \infty} \frac{L(-x)}{L(x)} = c \in (0, \infty).$$

Let

$$p = \frac{1}{2(1-\gamma)} \quad \text{and} \quad \theta_\lambda = \lambda L(\lambda^p)^{-2} \quad (\lambda > 0).$$

We introduce a rescaled process $(X^\lambda(t), B^\lambda(t))$ by

$$X_i^\lambda(t) = \lambda^{-p} X_i^\lambda(\theta_\lambda t), \quad B_i^\lambda(t) = \theta_\lambda^{-1/2} B_i(\theta_\lambda t), \quad i \in S.$$

Note that $\{B_i^\lambda(t)\}_{i \in S}$ are independent Brownians motion and the rescaled process $(X^\lambda(t), B^\lambda(t))$ satisfies the following SDE;

$$dX_i^\lambda(t) = \alpha_\lambda(X_i^\lambda(t)) dB_i^\lambda(t) + \theta_\lambda \sum_{j \in S} A_{ij}(X_j^\lambda(t) - X_i^\lambda(t)) dt,$$

where

$$\alpha_\lambda(x) = \lambda^{-p} \theta_\lambda^{1/2} \alpha(\lambda^p x).$$

Moreover it holds that

$$\lim_{\lambda \rightarrow \infty} \alpha_\lambda(x) = \begin{cases} x^\gamma & (0 < x), \\ c|x|^\gamma & (0 > x). \end{cases}$$

In order to describe the limiting processes of the $(X^\lambda(t))$ we introduce a class of skew Bessel processes on natural scale.

Let

$$\bar{\alpha}(x) = \begin{cases} \|m\|_2 x^\gamma & (0 \leq x), \\ \|m\|_2 c |x|^\gamma & (0 > x). \end{cases}$$

where $\|m\|_2 = \sqrt{\sum_{i \in S} m_i^2}$, $\bar{\alpha}(0) = \infty$ if $\gamma < 0$, and $\bar{\alpha}(0) = \|m\|_2$ if $\gamma = 0$.

Let us consider the following one-dimensional SDE:

$$(2.1) \quad \begin{aligned} dZ(t) &= \bar{\alpha}(Z(t)) dB(t), \\ Z(0) &= x \in \mathbb{R}. \end{aligned}$$

If $-\infty < \gamma \leq 0$, the SDE (2.1) has a law unique solution, however, if $0 < \gamma < 1/2$, the law uniqueness for (2.1) fails. In this case, if we add the non-sticky condition to (2.1), i.e.

$$(2.2) \quad \int_0^t I_{\{0\}}(Z(s)) ds = 0 \quad (\forall t > 0), \quad P\text{-a.s.},$$

the law uniqueness holds. In fact, the solution can be constructed from a Brownian motion through the time change method. Thus we have a diffusion process $(Z(t), P_x)$ on \mathbb{R} , which is called a *skew Bessel process on natural scale*.

Theorem 2.1. *Assume the conditions [B-1] and [B-2], and $X(0)$ is a \mathbb{R}^S -valued random variable independent of $B(t) = \{B_i(t)\}_{i \in S}$. Then*

$$(2.3) \quad (X^\lambda(t) = \{X_i^\lambda(t)\}_{i \in S}) \xrightarrow{(\mathcal{L})} (X^\infty(t) = \{X_i^\infty(t)\}_{i \in S}) \quad (\lambda \rightarrow \infty),$$

where $\xrightarrow{(\mathcal{L})}$ stands for the weak convergence of the probability laws on the path space induced by $\{X^\lambda(t)\}$. Moreover, all component processes of $\{X_i^\infty(t)\}_{i \in S}$ coincide with each other and the common process is equivalent to a skew Bessel diffusion on natural scale $(Z(t))$ governed by (2.1) with $Z(0) = 0$ being imposed the non-sticky condition whenever $0 < \gamma < 1/2$;

$$(2.4) \quad \int_0^t I_{\{0\}}(Z(s)) ds = 0 \quad (t > 0), \quad P\text{-a.s.}$$

From Theorem 2.1 it follows the following

Corollary 2.2. *Under the same assumption of in Theorem 2.1,*

$$(2.5) \quad X(t) \xrightarrow{(d)} q\delta_{+\infty} + (1 - q)\delta_{-\infty} \quad (t \rightarrow \infty).$$

Proof of Theorem 1.1. Theorem 1.1 follows immediately from Theorem 2.1. In fact, since

$$\int_0^t I(Z(s) = 0) ds = 0,$$

by Theorem 2.1 we can see that for every bounded continuous function f on $[-\infty, \infty]$ it holds that

$$\begin{aligned} \frac{1}{\theta_\lambda} \int_0^{\theta_\lambda} f(X(s)) ds &= \int_0^1 f(\lambda^p X^\lambda(s)) ds \\ &\stackrel{(d)}{\Rightarrow} f(+\infty) \int_0^1 I(Z(s) > 0) ds + f(-\infty) \int_0^1 I(Z(s) < 0) ds \\ &= Y_{p,q} f(+\infty) + (1 - Y_{p,q}) f(-\infty), \end{aligned}$$

because of

$$\int_0^1 I(Z(s) > 0) ds \stackrel{(d)}{=} Y_{p,q}.$$

For the last relation see Watanabe [5]. □

3. Proof of Theorem 2.1

To avoid complication of arguments we prove Theorem 2.1 under the following condition [B-3] instead of [B-1], since the proof is essentially the same even under the condition [B-1].

[B-3] Let $-\infty < \gamma < 1/2$, and for some $\alpha_+ > 0$ and $\alpha_- > 0$

$$(3.1) \quad \lim_{x \rightarrow \infty} \frac{\alpha(x)}{x^\gamma} = \alpha_+, \quad \lim_{x \rightarrow -\infty} \frac{\alpha(x)}{|x|^\gamma} = \alpha_-.$$

In what follows we assume the conditions [A-1], [A-2], [B-2] and [B-3]. Let

$$(3.2) \quad p = \frac{1}{2(1-\gamma)},$$

and for $\lambda > 0$ we set

$$(3.3) \quad \alpha_\lambda(x) = \lambda^{-p+1/2} \alpha(\lambda^p x),$$

and

$$(3.4) \quad \alpha_\infty(x) = \begin{cases} \alpha_+ x^\gamma & (0 \leq x), \\ \alpha_- |x|^\gamma & (x < 0). \end{cases}$$

where

$$\alpha_\infty(0) = \begin{cases} \infty & (\gamma < 0), \\ \alpha_+ & (\gamma = 0). \end{cases}$$

Moreover we set

$$(3.5) \quad \bar{\alpha}(x) = \|m\|_2 \alpha_\infty(x),$$

where $\{m_i\}_{i \in S}$ is a probability vector in (1.7), and $\|m\|_2 = \sqrt{\sum_{i \in S} m_i^2}$.

For the diffusion process $(X(t), P_x)$ governed by (1.1) we introduce a rescaled process $X^\lambda(t)$ ($\lambda > 0$) by

$$X_i^\lambda(t) = \lambda^{-p} X_i(\lambda t) \quad (i \in S),$$

which satisfies the following SDE:

$$(3.6) \quad dX_i^\lambda(t) = \alpha_\lambda(X_i^\lambda(t)) dB_i^\lambda(t) + \lambda \sum_{j \in S} A_{ij}(X_j^\lambda(t) - X_i^\lambda(t)) dt.$$

For the proof of Theorem 2.1 we may assume that the initial condition $X(0)$ is non-random, i.e.

$$X(0) = \{x_i\}_{i \in S} \in \mathbb{R}^S.$$

We first prepare several moment estimates of the rescaled process $X_i^\lambda(t)$.

Lemma 3.1. *Let $-\infty < \gamma < 1/2$. For $a \geq 2$ there exists a constant $C = C(a, p) > 0$ such that*

$$(3.7) \quad \sum_{i \in S} m_i E[|X_i^\lambda(t)|^a] \leq C \left(\lambda^{-pn} + \lambda^{-pn} \sum_{i \in S} m_i |x_i|^a + t^{pa} \right) \quad (t \geq 0, \lambda > 0).$$

Proof. Using the Itô formula and taking expectations, we have

$$(3.8) \quad \begin{aligned} \frac{d}{dt} \sum_{i \in S} m_i E[|X_i(t)|^a] &= a \sum_{i \in S} \sum_{j \in S} m_i A_{ij} E[|X_i(t)|^{a-1} \operatorname{sgn}(X_i(t))(X_j(t) - X_i(t))] \\ &\quad + \frac{1}{2} a(a-1) \sum_{i \in S} m_i E[|X_i(t)|^{a-2} \alpha^2(X_i(t))]. \end{aligned}$$

Note that

$$(3.9) \quad \sum_{i \in S} \sum_{j \in S} m_i A_{ij} |x_i|^{a-1} \operatorname{sgn}(x_i)(x_j - x_i) \leq 0,$$

because, using $\sum_{j \in S} A_{ij} = 0$, $\sum_{i \in S} m_i A_{ij} = 0$ and a simple inequality

$$t^{a-1}s \leq \frac{a-1}{a}t^a + \frac{1}{a}s^a \quad (t > 0, s > 0),$$

we see

$$\begin{aligned} & \sum_{i \in S} \sum_{j \in S} m_i A_{ij} |x_i|^{a-1} \operatorname{sgn}(x_i)(x_j - x_i) \\ & \leq \sum_{i \in S} \sum_{j \in S} m_i A_{ij} (|x_i|^{a-1} |x_j| - |x_i|^a) \\ & \leq \frac{1}{a} \sum_{i \in S} \sum_{j \in S} m_i A_{ij} (|x_j|^a - |x_i|^a) \\ & = 0. \end{aligned}$$

Note that by the conditions [A-1], [A-2] and [B-3] there exists constants $C_1 > 0$ and $C_2 > 0$ satisfying

$$(3.10) \quad C_1(1 + |x|)^\gamma \leq \alpha(x) \leq C_2(1 + |x|)^\gamma, \quad (x \in \mathbb{R}),$$

so that there exists a constant C_3 such that

$$(3.11) \quad \sum_{i \in S} m_i |x_i|^{a-2} \alpha^2(x_i) \leq C_3 \left(1 + \sum_{i \in S} m_i |x_i|^a \right)^{1-1/ap},$$

Hence, by (3.9), (3.10) and (3.11) $F(t) = \sum_{i \in S} m_i E[|X_i(t)|^{2a}]$ satisfies

$$\frac{d}{dt} F(t) \leq C_3(1 + F(t))^{1-1/ap}.$$

Thus we obtain, for some $C_4 > 0$,

$$(3.12) \quad \sum_{i \in S} m_i E[|X_i(t)|^a] \leq C_4 \left(1 + \sum_{i \in S} m_i |x_i|^a + t^{ap} \right).$$

(3.7) follows immediately from (3.12). □

Let

$$U_{i,j}(t) = X_i(t) - X_j(t) \quad (i \neq j \in S),$$

and for $\lambda > 0$ let

$$U_{i,j}^\lambda(t) = X_i^\lambda(t) - X_j^\lambda(t) \quad (i \neq j \in S).$$

Lemma 3.2. (i) For any $a \geq 2$ there exists a constant $C > 0$ such that

$$(3.13) \quad E[|U_{i,j}^\lambda(t)|^a] \leq \begin{cases} C\lambda^{-a/2}(1+t^{ap\gamma}) & (0 \leq \gamma < 1/2), \\ C\lambda^{-ap} & (-\infty < \gamma < 0). \end{cases}$$

(ii) For each $T > 0$ there exists a constant $C_T > 0$ such that for every $\lambda \geq 1$

$$(3.14) \quad E[|U_{i,j}^\lambda(t) - U_{i,j}^\lambda(s)|^6] \leq C_T\lambda^{-1}|t-s|^2 \quad (0 \leq s \leq t \leq T).$$

Proof. First, note that $X(t)$ satisfies

$$(3.15) \quad X_i(t) = \sum_{k \in S} \int_s^t Q_{t-u}(i, k) \alpha(X_k(u)) dB_k(u) + \sum_{k \in S} Q_{t-s}(i, j) X_k(s) \quad (i \in S),$$

so that

$$\begin{aligned} U_{i,j}(t) - U_{i,j}(s) &= \sum_{k \in S} \int_s^t (Q_{t-u}(i, k) - Q_{t-u}(j, k)) \alpha(X_k(u)) dB_k(u) \\ &\quad + \sum_{k \neq i} Q_{t-s}(i, k) U_{i,k}(s) + \sum_{k \neq j} Q_{t-s}(j, k) U_{j,k}(s). \end{aligned}$$

Using this and the Burkholder inequality, we have

$$(3.16) \quad \begin{aligned} &E[|U_{i,j}(t) - U_{i,j}(s)|^a] \\ &\leq C_1 \sum_{k \in S} E \left[\left(\int_s^t (Q_{t-u}(i, k) - Q_{t-u}(j, k))^2 \alpha^2(X_k(u)) du \right)^{a/2} \right] \\ &\quad + C_1 E \left[\left(\sum_{k \in S} Q_{t-s}(i, k) U_{i,k}(s) \right)^a \right] \\ &\quad + C_1 E \left[\left(\sum_{k \in S} Q_{t-s}(j, k) U_{j,k}(s) \right)^a \right]. \end{aligned}$$

When $0 \leq \gamma < 1/2$, using this with $s = 0$, (1.7) and Lemma 3.1 we have a constant $C_2 > 0$ satisfying that

$$E[|U_{i,j}(t)|^a] \leq C_2(1+t^{a\gamma p}),$$

which yields (3.13). Using (1.7), (3.10), and Lemma 3.1, we see that the first term of the r.h.s. of (3.16) with $a = 6$ is dominated by

$$C_3 \sum_{k \in S} E \left[\left(\int_s^t e^{-2\delta(t-u)} \alpha^2(X_k(u)) du \right)^3 \right] \leq C_4((t-s) \wedge 1)^3(1+t^{6p\gamma}).$$

Furthermore, by (3.13) the last two terms of (3.16) are dominated by

$$C_5((t - s) \wedge 1)^6(1 + t^{6p\gamma}),$$

thus we have

$$(3.17) \quad E[|U_{i,j}(t) - U_{i,j}(s)|^6] \leq C_6((t - s) \wedge 1)^3(1 + t^{6p\gamma}).$$

From this it follows that

$$\begin{aligned} E[|U_{i,j}^\lambda(t) - U_{i,j}^\lambda(s)|^6] &\leq C_7(\lambda(t - s) \wedge 1)^3\lambda^{-6p}(1 + (\lambda t)^{6p\gamma}) \\ &\leq C_T\lambda^{-1}(t - s)^2, \end{aligned}$$

which concludes (3.14). In the case $-\infty < \gamma < 0$, since $\alpha(x)$ is bounded, $E[U_{i,j}(t)^6]$ is also bounded in $t \geq 0$. Hence it is easy to obtain (3.14). \square

Lemma 3.3. *Suppose that $X(t)$ is a continuous martingale with $X(0) = 0$ defined on a complete probability space (Ω, \mathcal{F}, P) with filtration $\{\mathcal{F}_t\}$, of which quadratic variation process satisfies*

$$\langle X \rangle(t) = \int_0^t \bar{\alpha}^2(X(s)) ds,$$

where $\bar{\alpha}(x)$ is of (3.4). If $0 < \gamma < 1$, we further assume the non-sticky condition;

$$\int_0^t I_{(0)}(X(s)) ds = 0 \quad (t > 0) \quad P\text{-a.s.}$$

Then the probability law on the path space $W = C([0, \infty), \mathbb{R})$ induced by $(X(t))$ coincides with that of the skew Bessel process on natural scale $Z(t)$ starting at 0 governed by the SDE (2.1) with (2.2).

Proof. Proof is to verify that $X(t)$ satisfies the SDE (2.1) for some Brownian motion $\bar{B}(t)$ using the time-change method, that is quite standard, so we omit it. \square

Proof of Theorem 2.1 in case 0 $\gamma < 1/2$. In this case the proof is rather standard, that is, first to verify the tightness of the probability laws P^λ on W induced by $\{X^\lambda(t)\}$ and next to identify the limit of $\{P^\lambda\}$ as $\lambda \rightarrow \infty$.

For the stationary probability vector $\{m_i\}$ of Q_t we set

$$Y^\lambda(t) = \sum_{i \in S} m_i X_i^\lambda(t),$$

which satisfies the following equation;

$$(3.18) \quad dY^\lambda(t) = \sum_{i \in S} m_i \alpha_\lambda(X_i^\lambda(t)) dB_i^\lambda(t).$$

Lemma 3.4. *Let $0 \leq \gamma < 1/2$. For each $T > 0$ there exists constant $C_T > 0$ such that for every $\lambda > 0$,*

$$(3.19) \quad E[|Y^\lambda(t) - Y^\lambda(s)|^4] \leq C_T(t - s)^2, \quad (0 \leq s, t \leq T).$$

Proof. It is immediate from (3.18) and Lemma 3.1. □

Lemma 3.5. *Let $0 \leq \gamma < 1/2$.*

$$(3.20) \quad \lim_{\varepsilon \rightarrow 0^+} \limsup_{\lambda \rightarrow \infty} \int_0^t P(|X_i^\lambda(s)| \leq \varepsilon) ds = 0. \quad (i \in S, t > 0).$$

Proof. For each $\varepsilon > 0$ define a function φ_ε by

$$\begin{aligned} \varphi_\varepsilon''(x) &= |x|^{-2\gamma} I(|x| \leq \varepsilon), \\ \varphi_\varepsilon(x) &= \int_0^{|x|} \int_0^y \varphi_\varepsilon''(u) du dy. \end{aligned}$$

Applying Itô formula we obtain

$$(3.21) \quad E[\varphi_\varepsilon(Y^\lambda(t))] = \varphi_\varepsilon\left(\lambda^{-p} \sum_{i \in S} m_i x_i\right) + \sum_{i \in S} \int_0^t m_i^2 E[\alpha_\lambda^2(X_i^\lambda(s)) \varphi_\varepsilon''(Y^\lambda(s))] ds.$$

Since

$$|\varphi_\varepsilon(x)| \leq \frac{\varepsilon^{1-2\gamma}}{(1-2\gamma)} |x|,$$

using Lemma 3.1 we have

$$(3.22) \quad \lim_{\varepsilon \rightarrow 0^+} \limsup_{\lambda \rightarrow \infty} \sum_{i \in S} \int_0^t m_j^2 E[\alpha_\lambda^2(X_i^\lambda(s)) \varphi_\varepsilon''(Y^\lambda(s))] ds = 0.$$

Note that for some $C_1 > 0$

$$\alpha_\lambda^2(x) \geq C_1(\lambda^{-p} + |x|)^\gamma \quad (x \in \mathbb{R}, \lambda > 0),$$

and for $y = \sum_i m_i x_i$

$$\begin{aligned} \sum_{i \in S} m_i \alpha_\lambda^2(x_i) \varphi_\varepsilon''(y) &\geq C_1 \sum_{i \in S} m_i (\lambda^{-p} + |x_i|)^{2\gamma} |y|^{-2\gamma} I(|y| < \varepsilon) \\ &\geq C_2 (\lambda^{-p} + |y|)^{2\gamma} |y|^{-2\gamma} I(|y| < \varepsilon) \\ &\geq C_2 I(|y| < \varepsilon). \end{aligned}$$

Hence from this and (3.22) it follows that

$$(3.23) \quad \lim_{\varepsilon \rightarrow 0^+} \limsup_{\lambda \rightarrow \infty} \int_0^t P(|Y^\lambda(s)| \leq \varepsilon) du = 0.$$

Here we notice that

$$P(|X_i^\lambda(s)| \leq \varepsilon) \leq P(|Y^\lambda(s)| \leq 2\varepsilon) + P(|X_i^\lambda(s) - Y^\lambda(s)| > \varepsilon),$$

and that for each $\varepsilon > 0$ the second term vanishes as $\lambda \rightarrow \infty$. Hence (3.20) follows from (3.23). □

Now we proceed to the proof of Theorem 2.1 in the case $0 \leq \gamma < 1/2$. Let P^λ be the probability measure on $W = C([0, \infty), \mathbb{R}^S)$ induced by $X^\lambda(t)$. We use the notation E^{P^λ} for the expectation by P^λ . Then by Lemma 3.4 and Lemma 3.2 $\{P^\lambda\}$ is tight. Suppose that for some $\{\lambda_n\}$ tending to ∞ , P^{λ_n} converges weakly to P^∞ . Let

$$\bar{w}(t) = \sum_{i \in S} m_i w_i(t).$$

Since by (3.18) $\bar{w}(t)$ is a P^λ -martingale with quadratic variation process

$$(3.24) \quad \langle \bar{w} \rangle(t) = \sum_{i \in S} m_i^2 \int_0^t \alpha_\lambda^2(w_i(s)) ds \quad P^\lambda\text{-a.s.},$$

using Lemma 3.1 we see easily that $\bar{w}(t)$ is a P^∞ -martingale with $\bar{w}(0) = 0$. Moreover, it follows from Lemma 3.2 that

$$(3.25) \quad P^\infty(w_i(t) = w_j(t) \ (\forall t \geq 0)) = 1.$$

(3.24) implies that for every $0 \leq s < t$ and a \mathcal{F}_s -measurable and bounded continuous function $\Phi_s(w)$ on W

$$(3.26) \quad E^{P^\lambda} \left[\left(\bar{w}^2(t) - \bar{w}^2(s) - \sum_{i \in S} m_i^2 \int_s^t \alpha_\lambda^2(w_i(u)) du \right) \Phi_s(w) \right] = 0.$$

We claim that

$$(3.27) \quad \lim_{\lambda \rightarrow \infty} E^{P^\lambda} \left[\left(\int_s^t \alpha_\lambda^2(w_i(u)) du \right) \Phi_s(w) \right] = E^{P^\infty} \left[\left(\int_s^t \bar{\alpha}^2(w_i(u)) du \right) \Phi_s(w) \right].$$

For $\varepsilon > 0$ let φ_ε be a smooth function on \mathbb{R} satisfying

$$I_{\mathbb{R} \setminus [-\varepsilon, \varepsilon]}(x) \leq \varphi_\varepsilon(x) \leq I_{\mathbb{R} \setminus [-\varepsilon/2, \varepsilon/2]}(x).$$

Since $\alpha_\lambda(x)$ converges to $\alpha_\infty(x)$ as $\lambda \rightarrow \infty$ compact uniformly in $\mathbb{R} \setminus \{0\}$ and

$$\alpha_\lambda(x) \leq C_3(1 + |x|^\nu) \quad (x \in \mathbb{R}),$$

using Lemma 3.1 we see that for every $\varepsilon > 0$

$$\begin{aligned} (3.28) \quad & \lim_{\lambda \rightarrow \infty} E^{P^\lambda} \left[\left(\int_s^t \alpha_\lambda^2(w_i(u)) \varphi_\varepsilon(w_i(u)) \, du \right) \Phi_s(w) \right] \\ & = E^{P^\infty} \left[\left(\int_s^t \alpha_\infty^2(w_i(u)) \varphi_\varepsilon(w_i(u)) \, du \right) \Phi_s(w) \right]. \end{aligned}$$

On the other hand by Lemma 3.5

$$\begin{aligned} & \lim_{\varepsilon \rightarrow +0} \limsup_{\lambda \rightarrow \infty} E^{P^\lambda} \left[\left(\int_s^t \alpha_\lambda^2(w_i(u))(1 - \varphi_\varepsilon)(w_i(u)) \, du \right) \Phi_s(w) \right] \\ & \leq C_4 \lim_{\varepsilon \rightarrow +0} \limsup_{\lambda \rightarrow \infty} \int_0^t P(|X_i^\lambda(u)| \leq \varepsilon) \, du = 0. \end{aligned}$$

(3.27) follows from this and (3.28). Thus, $\bar{w}(t)$ is a P^∞ -martingale with quadratic variation process

$$\langle w \rangle(t) = \sum_{i \in S} m_i^2 \int_0^t \alpha_\infty^2(w_i(u)) \, du = \int_0^t \bar{\alpha}^2(\bar{w}(u)) \, du.$$

Therefore by Lemma 3.3 P^∞ coincides with the probability law of the skew Bessel process on natural scale, which completes the proof of Theorem 2.1 in the case $0 \leq \gamma < 1/2$. □

Proof of Theorem 2.1 in case $\gamma < 0$. In this case it seems hard to obtain the moment estimate for $Y^\lambda(t)$ as in Lemma 3.4 due to difficulty of negative power moment estimates, so we consider a spatial transformation by an asymptotic scale function $S(x)$;

$$S(x) = \begin{cases} x^{2(1-\gamma)} & (\gamma \geq 0), \\ |x|^{2(1-\gamma)} & (\gamma < 0). \end{cases}$$

Lemma 3.6. *Let $-\infty < \gamma < 0$. For each $T > 0$ there exists a constant $C_T > 0$ such that for every $\lambda \geq 1$*

$$(3.29) \quad E[|S(Y^\lambda(t)) - S(Y^\lambda(s))|^4] \leq C_T |t - s|^2, \quad (0 \leq s, t \leq T).$$

Proof. Recall that $Y^\lambda(t)$ satisfies

$$(3.30) \quad dY^\lambda(t) = \alpha_\lambda(Y^\lambda(t)) \, dV^\lambda(t),$$

where $V^\lambda(t)$ is a continuous martingale with quadratic variation process

$$(3.31) \quad \langle V^\lambda \rangle(t) = \sum_{i \in S} m_i^2 \int_0^t \frac{\alpha_\lambda^2(X_i^\lambda(u))}{\alpha_\lambda^2(Y^\lambda(u))} du.$$

Applying Itô formula to $S(x)$ together with Burkholder's inequality we see that

$$(3.32) \quad \begin{aligned} & E[|S(Y^\lambda(t)) - S(Y^\lambda(s))|^4] \\ & \leq C_1 E \left[\left(\int_s^t |S'(Y^\lambda(u))|^2 \alpha_\lambda^2(Y^\lambda(u)) d\langle V^\lambda \rangle(u) \right)^2 \right] \\ & \quad + C_1 E \left[\left(\int_s^t S''(Y^\lambda(u)) \alpha_\lambda^2(Y^\lambda(u)) d\langle V^\lambda \rangle(u) \right)^4 \right] \\ & \leq C_1 \int_s^t E[(S' \alpha_\lambda)^4(Y^\lambda(u))] du \int_s^t E[(\langle V^\lambda \rangle'(u))^4] du \\ & \quad + C_1 \|S'' \alpha_\lambda^2\|_\infty^4 (t-s)^3 \int_s^t E[(\langle V^\lambda \rangle'(u))^4] du, \end{aligned}$$

where

$$\langle V^\lambda \rangle'(u) = \sum_{i \in S} m_i^2 \frac{\alpha_\lambda^2(X_i^\lambda(u))}{\alpha_\lambda^2(Y^\lambda(u))},$$

and we notice that $S'' \alpha_\lambda^2(x)$ is bounded in $x \in \mathbb{R}$ and $\lambda \geq 1$. Note that

$$C_2 \lambda^{2p-1} (1 + \lambda^p |x|)^{2\gamma} \leq \alpha_\lambda^2(x) \leq C_3 \lambda^{2p-1} (1 + \lambda^p |x|)^{2\gamma},$$

then

$$\frac{\alpha_\lambda^2(x)}{\alpha_\lambda^2(y)} \leq C_4 \left(\frac{1 + \lambda^p |y|}{1 + \lambda^p |x|} \right)^{2|\gamma|} \leq C_4 (1 + \lambda^p |x - y|)^{2|\gamma|}.$$

Hence,

$$E[(\langle V^\lambda \rangle'(u))^4] \leq C_5 \left(1 + \lambda^{8p|\gamma|} \sum_{j \neq k} E[|U_{j,k}^\lambda(u)|^{8|\gamma|}] \right),$$

which is bounded in $u \geq 0$ by Lemma 3.2. Accordingly, it follows from this and (3.32) that

$$E[(S(Y^\lambda(t)) - S(Y^\lambda(s)))^4] \leq C_7 (|t - s|^2 + |t - s|^4),$$

which completes the proof of Lemma 3.6. □

Lemma 3.7. *Let $-\infty < \gamma < 0$. Then*

$$(3.33) \quad \lim_{\varepsilon \rightarrow 0} \limsup_{\lambda \rightarrow \infty} \int_0^t E[\alpha_\lambda^2(X_i^\lambda(s))I(|X_i^\lambda(s)| \leq \varepsilon)] ds = 0 \quad (i \in S).$$

Proof. In the proof of Lemma 3.5, replacing $\varphi_\varepsilon(x)$ by $\varphi_\varepsilon''(x) = I_{[-\varepsilon, \varepsilon]}(x)$ we have

$$(3.34) \quad \lim_{\varepsilon \rightarrow 0^+} \limsup_{\lambda \rightarrow \infty} \sum_{i \in S} \int_0^t m_i^2 E[\alpha_\lambda^2(X_i^\lambda(s))I_{[-\varepsilon, \varepsilon]}(Y^\lambda(s)) ds] = 0.$$

Noting that

$$I_{[-\varepsilon, \varepsilon]}(X_i^\lambda(s)) \leq I_{[-2\varepsilon, 2\varepsilon]}(Y^\lambda(s)) + \sum_{j \in S} I_{[-\varepsilon, \varepsilon]}(X_j^\lambda(s) - X_i^\lambda(s)),$$

and by Lemma 3.2 we can see

$$\begin{aligned} & \lim_{\lambda \rightarrow \infty} \int_0^t E[\alpha_\lambda^2(X_i(s))I_{[-\varepsilon, \varepsilon]}(X_j^\lambda(s) - X_i^\lambda(s))] ds \\ & \leq \lim_{\lambda \rightarrow \infty} \lambda^{-2p+1} \|\alpha\|_\infty^2 \int_0^t P(|U_{i,j}^\lambda(s)| > \varepsilon) = 0. \end{aligned}$$

Thus (3.33) follows from this and (3.34). □

Now we are in position to complete the proof of Theorem 2.1 in the case $-\infty < \gamma < 0$, but one can proceed the proof as in the case of $0 \leq \gamma < 1/2$, so we shall only sketch the proof. By virtue of Lemma 3.2 and Lemma 3.6, we may assume that P^{λ_n} converges weakly to P^∞ as $n \rightarrow \infty$ for some $\lambda_n \nearrow \infty$. Then, $\bar{w}(t)$ is P^∞ -martingale with $\bar{w}(0) = 0$ and

$$w_i(t) = w_j(t) = \bar{w}(t) \quad P^\infty\text{-a.s.} \quad (i, j \in S),$$

in the same way as $0 \leq \gamma < 1/2$. Using the Lemma 3.7 instead of Lemma 3.5, we also have (3.27) which implies

$$E^{P^\infty} \left[\left(\bar{w}^2(t) - \bar{w}^2(s) - \int_s^t \bar{\alpha}^2(\bar{w}(u)) du \right) \Phi_s(w) \right] = 0,$$

and then by Lemma 3.3, the probability law $(\bar{w}(t), P^\infty)$ coincides with that of the desired skew Bessel process on natural scale. Therefore Theorem 2.1 has been proved completely. □

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References

- [1] I. Karatzas and S.E. Shreve: *Brownian Motion and Stochastic Calculus*, Second edition, Springer, New York, 1991.
- [2] J. Lamperti: *An occupation time theorem for a class of stochastic processes*, Trans. Amer. Math. Soc. **88** (1958), 380–387.
- [3] D. Revuz and M. Yor: *Continuous Martingales and Brownian Motion*, Springer, Berlin, 1991.
- [4] D.W. Stroock and S.R.S. Varadhan: *Multidimensional Diffusion Processes*, Springer, Berlin, 1979.
- [5] S. Watanabe: *Generalized arc-sin laws for one-dimensional diffusion processes and random walks*; in *Stochastic Analysis* (Ithaca, NY, 1993), Proc. Sympos. Pure Math. **57**, Amer. Math. Soc., Providence, RI, 1995, 157–172.

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