

# LARGE DEVIATIONS FOR RANDOM WALKS WITH TIME STATIONARY RANDOM DISTRIBUTION FUNCTION

DUG HUN HONG

## 1. Introduction

Let  $\mathcal{F}$  be a set of distributions on  $R$  with the topology of weak convergence, and let  $\mathcal{A}$  be the  $\sigma$ -field generated by the open sets. We denote by  $\mathcal{F}_1^\infty$  the space consisting of all infinite sequence  $(F_1, F_2, \dots)$ ,  $F_n \in \mathcal{F}$  and  $R_1^\infty$  the space consisting of all infinite sequences  $(x_1, x_2, \dots)$  of real numbers. Take the  $\sigma$ -field  $\mathcal{A}_1^\infty$  to be the smallest  $\sigma$ -field of subsets of  $\mathcal{F}_1^\infty$  containing all finite-dimensional rectangles and take  $\mathcal{B}_1^\infty$  to be the Borel  $\sigma$ -field of  $R_1^\infty$ . Let  $\omega = (F_1^\omega, F_2^\omega, \dots)$  be the coordinate process in  $R_1^\infty$  and  $\nu$  its distribution on  $\mathcal{A}_1^\infty$ . Let  $\theta$  be the coordinate shift :  $\theta^k(\omega) = \omega'$  with  $F_n^{\omega'} = F_{n+k}^\omega$ ,  $k = 1, 2, \dots$ . On  $(R_1^\infty, \mathcal{B}_1^\infty)$  we also define the shift transformation  $\sigma : R_1^\infty \rightarrow R_1^\infty$  by  $\sigma(x_1, x_2, \dots) = (x_2, x_3, \dots)$ .  $\nu$  is called stationary if for every  $A \in \mathcal{A}_1^\infty$ ,  $\nu(\theta^{-1}(A)) = \nu(A)$  and we let  $\pi$  be its marginal distribution. Let  $\mathcal{I}$  be the  $\sigma$ -field of invariant sets in  $\mathcal{A}_1^\infty$ , that is,  $\mathcal{I} = \{A | \theta^{-1}(A) = A, A \in \mathcal{A}_1^\infty\}$  and let  $\mathcal{J}$  be the  $\sigma$ -field of invariant sets in  $\mathcal{B}_1^\infty$ , that is,  $\mathcal{J} = \{B | \sigma^{-1}(B) = B, B \in \mathcal{B}_1^\infty\}$ .  $\nu$  is called independent and identically distributed (i.i.d.) if  $\nu$  is stationary and product measure. For each  $\omega$ , define a probability measure  $P_\omega$  on  $(R_1^\infty, \mathcal{B}_1^\infty)$  so that  $P_\omega = \prod_{i=1}^\infty F_i^\omega$ . A monotone class argument shows that  $P_\omega(B)$ ,  $B \in \mathcal{B}_1^\infty$ , is  $\mathcal{A}_1^\infty$ -measurable as a function of  $\omega$ . So we can define a new probability measure such that  $P(B) = \int P_\omega(B) \nu(d\omega)$ . Define the process  $\{X_n\}$  on  $(R_1^\infty, \mathcal{B}_1^\infty)$  such that  $X_n(x_1, x_2, \dots) = x_n$  and set  $S_n = X_1 + X_2 + \dots + X_n$ . By the definition of  $P_\omega$ ,  $\{X_n\}$  are independent with respect to  $P_\omega$  and we also note that  $\{X_n\}$  is a sequence of independent and identically distributed random variables

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sequence of independent and identically distributed random variables when  $\mathcal{F}$  has just one element. In this paper we generalize the classical Cramer theorem in this set up.

### 2. Strong law of large numbers

In this section we consider some strong law of large numbers which are used to prove the main results.

**LEMMA 1.** *Let  $\mathcal{F} = \{F | \int |x|dF(x) < \infty, \int xdF(x) = 0\}$ , and let  $\nu$  be stationary with  $\int \int |x|dF(x)\pi(dF) < \infty$ . Then  $X_1$  with respect to  $P$  satisfies*

$$E[X_1 | \mathcal{J}] = 0 \quad \text{a.s.}$$

*Proof.* By the assumption,  $E|X_1| < \infty$  and hence  $E[X_1 | \mathcal{J}]$  exists. Now let  $A \in \mathcal{J}$  and let  $\{(X_1, X_2, \dots) \in B\} = A$  for some  $B \in \mathcal{B}_1^\infty$ . Then we have

$$\begin{aligned} \int_A X_1 dP &= \int_{\{(X_1, X_2, \dots) \in B\}} X_1 dP \\ &= \int_{\{(X_2, X_3, \dots) \in B\}} X_1 dP \\ &= \int \left( \int x_1 dF_1^\omega(x_1) \int_B \prod_{i=2}^\infty dF_i^\omega(x_i) \right) \nu(d\omega) \\ &= 0, \end{aligned}$$

where the last equality holds because  $\int xdF(x) = 0$  for all  $F \in \mathcal{F}$ . This proves the lemma.

**THEOREM 1.** *Let  $\mathcal{F} = \{F | \int xdF(x) = 0, \int |x|dF(x) < \infty\}$  and  $\nu$  be stationary with  $\int \int |x|dF(x)\pi(dF) < \infty$ . Then*

$$P_\omega \left\{ \frac{S_n}{n} \rightarrow 0 \right\} = 1, \quad \nu - \text{a.e. } \omega.$$

*Proof.* The proof follows directly from Proposition 1 and 3[5], Lemma 1, and the Birkhoff's ergodic theorem.

In general we then prove the following theorem.

**THEOREM 2.** *Let  $\mathcal{F} = \{F \mid \int |x|dF(x) < \infty\}$  and let  $\nu$  be stationary with  $\int \int |x|dF(x)\pi(dF) < \infty$ . Then*

$$P_\omega \left\{ \frac{S_n}{n} \rightarrow E \left[ \int x dF_1(x) | \mathcal{I} \right] (\omega) \right\} = 1, \quad \nu - \text{a.e. } \omega.$$

( $E[\int x dF_1(x) | \mathcal{I}](\omega) = E[\int x dF_1^\omega(x)] = \int \int x dF(x)\pi(dF)$  in case  $\nu$  is ergodic.)

*Proof.* By Theorem 1,  $P_\omega \left\{ \frac{S_n - E_\omega S_n}{n} \rightarrow 0 \right\} = 1$ ,  $\nu$ -a.e.  $\omega$ , where  $E_\omega S_n = \sum_{k=1}^n \int X_k dP_\omega = \sum_{k=1}^n \int x dF_k^\omega(x)$ . We know  $\frac{1}{n} E_\omega S_n \rightarrow E[\int x dF_1(x) | \mathcal{I}](\omega)$ ,  $\nu$ -a.e.  $\omega$  by the ergodic theorem. Hence

$$P_\omega \left\{ \frac{S_n}{n} \rightarrow E \left[ \int x dF_1(x) | \mathcal{I} \right] (\omega) \right\} = 1, \quad \nu - \text{a.e. } \omega.$$

### 3. Large deviations

We begin this section by introducing the logarithmic moment generating function  $C_F(\xi) = \log M_F(\xi)$  where  $M_F(\xi) = \int \exp(\xi x) dF(x)$ ,  $\xi \in R$ , and  $C(\xi) = \int_{\mathcal{F}} C_F(\xi) \pi(dF)$ ,  $\xi \in R$ . Throughout this section we assume

$$(3.1) \quad M_F(\xi) < \infty \text{ for all } F \in \mathcal{F} \text{ and for all } \xi \in R,$$

$$(3.2) \quad C(\xi) < \infty \text{ for all } \xi \in R.$$

Note that since  $\xi \in R \rightarrow C_F(\xi)$  is a convex function, for each  $F \in \mathcal{F}$ , so is  $C(\xi)$ . Next let  $K(x)$  be the Legendre transform of  $C(\xi)$  :

$$(3.3) \quad K(x) \equiv \sup \{ \xi x - C(\xi) | \xi \in R \}, \quad x \in R.$$

Note that, by its definition as the pointwise supremum of linear functions,  $K(x)$  is necessarily a convex function. In order to develop some feeling for the relationship between  $C(\xi)$  and  $K(x)$ , we present the following elementary lemma.

LEMMA 2.  $K(x) \geq 0$ , moreover, if  $\int \int |x|dF(x)\pi(dF) < \infty$  and  $p = \int \int x dF(x)\pi(dF)$  then  $K(p) = 0$ ,  $K$  is non-decreasing on  $[p, \infty)$  and non-increasing on  $(-\infty, p]$ . In addition, for  $q \geq p$ ,  $K(q) = \sup\{\xi q - C(\xi) | \xi \geq 0\}$  and for  $q \leq p$ ,  $K(q) = \sup\{\xi q - C(\xi) | \xi \leq 0\}$ .

*Proof.* We begin by noting that, since  $\xi x - C(\xi) = 0$  for  $\xi = 0$  and for every  $x \in R$ ,  $K(\xi) \geq 0$ . Now suppose that  $\int \int |x|dF(x)\pi(dF) < \infty$  and set  $p = \int \int x dF(x)\pi(dF)$ . To see that  $K(p) = 0$ , we use Jensen's inequality to obtain

$$\begin{aligned} C(\xi) &= \int_{\mathcal{F}} (\log \int \exp(\xi x) dF(x)) \pi(dF) \\ &\geq \int_{\mathcal{F}} \int \xi x dF(x) \pi(dF) = \xi p \quad \text{for all } \xi \in R. \end{aligned}$$

In particular, this shows that  $\xi p - C(\xi) \leq 0$  for all  $\xi \in R$  and hence  $K(p) \leq 0$ . Since  $K(x)$  is non-negative and convex, this proves that  $K(p) = 0$ , that  $K(x)$  is non-decreasing on  $[p, \infty)$ , and that  $K(x)$  is non-increasing on  $(-\infty, p]$ .

As a consequence of Lemma 2, we have the following.

LEMMA 3. Let  $\mathcal{F} = \{F | \int \exp(\xi x) dF(x) < \infty, \xi \in R\}$ . If  $\nu$  is stationary and ergodic with  $\int \int |x|dF(x)\pi(dF) < \infty$ , then for every closed set  $G \subset R$ ,

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log P_{\omega} \left\{ \frac{S_n}{n} \in G \right\} \leq -\inf\{K(x) | x \in G\}, \quad \nu - a.e. \omega.$$

*Proof.* Let  $p = \int \int x dF(x)\pi(dF)$ . Suppose  $q \geq p$  ( $q \leq p$ ). For  $\xi \geq 0$ ,

$$\begin{aligned} P_{\omega} \left\{ \frac{S_n}{n} \geq q \right\} &\leq \exp(-\xi q) E_{\omega} \exp\left(\xi \frac{S_n}{n}\right) \\ &= \exp(-\xi q) \Pi_{i=1}^n E_{\omega} \exp\left(\xi \frac{X_i}{n}\right). \end{aligned}$$

Then

$$\begin{aligned} \frac{1}{n} \log P_{\omega} \left\{ \frac{S_n}{n} \geq q \right\} &\leq \frac{1}{n} \log(\exp(-\xi q) \Pi_{i=1}^n E_{\omega} \left(\exp\left(\xi \frac{X_i}{n}\right)\right)) \\ &= -\frac{\xi}{n} q + \frac{1}{n} \sum_{i=1}^n \log \int \exp\left(\frac{\xi x}{n}\right) dF_i^{\omega}(x). \end{aligned}$$

Note that

$$\frac{1}{n} \sum_{i=1}^n \log \int \exp(\xi x) dF_i^\omega(x) \rightarrow \int (\log \int \exp(\xi x) dF(x)) \pi(dF) = C(\xi)$$

$\nu$ -a.e.  $\omega$  by the ergodic theorem. Then for given  $\epsilon > 0$ , and  $\nu$ -a.e.  $\omega$  that we have for  $n \geq N(\omega)$

$$\frac{1}{n} \log P_\omega \left\{ \frac{S_n}{n} \geq q \right\} \leq \inf \{ -\xi q + C(\xi) \mid \xi \geq 0 \} + \epsilon.$$

Since  $\epsilon$  is arbitrary, by Lemma 2,

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log P_\omega \left\{ \frac{S_n}{n} \geq q \right\} \leq \inf \{ -\xi q + C(q) \mid \xi \geq 0 \} = -K(q).$$

Since  $K$  is non-decreasing (non-increasing) on  $[p, \infty)$  (on  $(-\infty, p]$ ) above inequality proves the result when either  $G \subset [p, \infty)$  or  $G \subset (-\infty, p]$ . On the other hand, if both  $G \cap [p, \infty) \neq \phi$  and  $G \cap (-\infty, p] \neq \phi$ , let  $q_+ = \inf \{ x \geq p \mid x \in G \}$  and  $q_- = \sup \{ x \leq p \mid x \in G \}$ . Then for  $\xi_1 \geq 0, \xi_2 \geq 0$

$$\begin{aligned} & P_\omega \left\{ \frac{S_n}{n} \in G \right\} \\ & \leq \exp(-\xi_1 q_+) E_\omega \left( \exp \left( \xi_1 \frac{S_n}{n} \right) \right) + \exp(\xi_2 q_-) E_\omega \left( \exp \left( -\xi_2 \frac{S_n}{n} \right) \right) \end{aligned}$$

and hence

$$\begin{aligned} & \limsup_{n \rightarrow \infty} \frac{1}{n} \log P_\omega \left\{ \frac{S_n}{n} \in G \right\} \\ & \leq \max \left[ \limsup_{n \rightarrow \infty} \frac{1}{n} \log \left( \exp(-\xi_1 q_+) E_\omega \left( \exp \left( \xi_1 \frac{S_n}{n} \right) \right) \right), \right. \\ & \quad \left. \limsup_{n \rightarrow \infty} \frac{1}{n} \log \left( \exp(\xi_2 q_-) E_\omega \left( \exp \left( -\xi_2 \frac{S_n}{n} \right) \right) \right) \right] \\ & \leq \max[-K(q_+), -K(q_-)] = -\inf \{ K(x) \mid x \in G \}. \end{aligned}$$

For the lower bound we need the following lemma. We define  $F^{-1}(t) = \sup \{ x \mid F(x) \leq t \}$ ,  $t \in (0, 1)$ .

LEMMA 4. Suppose that for every  $F \in \mathcal{F}$ ,  $F^{-1}$  is unbounded below and above and that there exists a measurable function  $\phi(\xi, F)$  such that

$$(3.4) \quad \left| \int \frac{x \exp(\xi x)}{M_F(\xi)} dF(x) \right| \leq \phi(\xi, F),$$

$$(3.5) \quad \int \left( \sup_{|\xi| \leq \xi_0} \phi(\xi, F) \right) \pi(dF) < \infty \quad \text{for all } \xi_0 \in R.$$

Then we have

- i)  $f(\xi) = \int \int \frac{x \exp(\xi x)}{M_F(\xi)} dF(x) \pi(dF)$  is continuous and  $\lim_{\xi \rightarrow \pm\infty} f(\xi) = \pm\infty$ ,
- ii) For each  $q$ ,  $K(q) = \sup\{\xi q - c(\xi) | \xi \in R\}$  is assumed at some point  $\xi = \xi_c(q)$  or equivalently  $C'(\xi_c(q)) = q$ .

*Proof.* i) We know that for all  $F$ , the function  $\xi \rightarrow \int \frac{x \exp(\xi x)}{M_F(\xi)} dF(x)$  is continuous and  $\int \frac{x \exp(\xi x)}{M_F(\xi)} dF(x) \rightarrow +\infty(-\infty)$  as  $\xi \rightarrow +\infty(-\infty)$  by unboundedness of  $F^{-1}$ . So  $f(\xi)$  is continuous by the Lebesgue dominated convergence theorem using (3.4) and (3.5) and we can easily check  $f(\xi) \rightarrow +\infty(-\infty)$  as  $\xi \rightarrow +\infty(-\infty)$ .

ii) consider the following:

$$\begin{aligned} & \lim_{\xi \rightarrow \xi'} \frac{C(\xi) - C(\xi')}{\xi - \xi'} \\ &= \lim_{\xi \rightarrow \xi'} \int \frac{C_F(\xi) - C_F(\xi')}{\xi - \xi'} \pi(dF) \\ &= \lim_{\xi \rightarrow \xi'} \int C'_F(\xi'') \pi(dF), \quad \text{where } \xi'' \in (\xi, \xi') \quad \text{or } \xi'' \in (\xi', \xi) \\ &= \lim_{\xi \rightarrow \xi'} \int \int \frac{x \exp(\xi'' x)}{M_F(\xi'')} dF(x) \pi(dF) \\ &= \int \lim_{\xi \rightarrow \xi'} \int \frac{x \exp(\xi'' x)}{M_F(\xi'')} dF(x) \pi(dF) \\ &= \int \int \frac{x \exp(\xi' x)}{M_F(\xi')} dF(x) \pi(dF) = f(\xi'); \end{aligned}$$

hence  $C'(\xi) = f(\xi)$ .

The fourth equality above follows from (3.4), (3.5) and the dominated convergence theorem. Note that  $C$  is convex. So for every given  $q$  there exists  $\xi_c(q)$  such that  $C'(\xi_c(q)) = q$ . This proves the lemma.

**THEOREM 3.** *Suppose that  $\nu$  is stationary and ergodic. Then under (3.1), (3.2), (3.4) and (3.5) for every measurable  $\Gamma \subset R$  we have that*

$$\begin{aligned} & - \inf \{K(x) | x \in \Gamma^o\} \\ \leq & \liminf_{n \rightarrow \infty} \frac{1}{n} \log P_\omega \left( \frac{S_n}{n} \in \Gamma \right) \leq \limsup_{n \rightarrow \infty} \frac{1}{n} \log P_\omega \left( \frac{S_n}{n} \in \Gamma \right) \\ \leq & - \inf \{K(x) | x \in \bar{\Gamma}\} \quad \nu - a.e. \omega. \end{aligned}$$

*Proof.* In view of Lemma 3, we only need to show that if  $q \in R$  and  $\delta > 0$ ,

$$(3.6) \quad \liminf_{n \rightarrow \infty} \frac{1}{n} \log P_\omega \left( \frac{S_n}{n} \in (q - \epsilon, q + \epsilon) \right) \geq -K(q).$$

In proving (3.6), we first suppose that for all  $F \in \mathcal{F}$ ,  $F^{-1}$  is unbounded above and below. Then for each  $q$ , there exists  $\xi$  such that  $C'(\xi) = f(\xi) = q$  by Lemma 4, and so  $K(q) = \xi q - C(\xi)$ .

$$\begin{aligned} & P_\omega \left\{ \frac{S_n}{n} \in (q - \delta, q + \delta) \right\} \\ = & \int \left\{ \frac{x_1 + \dots + x_m}{n} \in (q - \delta, q + \delta) \right\} dF_1^\omega(x_1) \dots dF_n^\omega(x_n) \\ \geq & M_{F_1^\omega}(\xi) \dots M_{F_n^\omega}(\xi) \exp(-\xi(q + \delta)n) \\ & \times \int \left\{ \frac{x_1 + \dots + x_n}{n} \in (q - \delta, q + \delta) \right\} \\ & \frac{\exp(\xi x_1)}{M_{F_1^\omega}(\xi)} dF_1^\omega(x_1) \dots \frac{\exp(\xi x_n)}{M_{F_n^\omega}(\xi)} dF_n^\omega(x_n). \end{aligned}$$

Here we need the following lemma.

LEMMA 5. Under the conditions of Theorem 3, we have for  $\nu$ -a.e.  $\omega$

$$\int \left\{ \frac{x_1 + \dots + x_n}{n} \in (q - \delta, q + \delta) \right\} \frac{\exp(\xi x_1)}{M_{F_1^\omega}(\xi)} dF_1^\omega(x_1) \dots \frac{\exp(\xi x_n)}{M_{F_n^\omega}(\xi)} dF_n^\omega(x_n) \rightarrow 1,$$

as  $n \rightarrow \infty$ .

*Proof.* For given  $\xi$  define  $\dot{F}$  so that  $\dot{F}(t) = \int_{-\infty}^t \frac{\exp(\xi x)}{M_F(\xi)} dF(x)$ . Let  $\dot{\mathcal{F}} = \{\dot{F} | F \in \mathcal{F}\}$ . Define  $\phi : \mathcal{F}_1^\infty \rightarrow \dot{\mathcal{F}}_1^\infty$  by  $\phi(\omega) = \dot{\omega} = (\dot{F}_1^\omega, \dot{F}_2^\omega, \dots)$ . Now let  $\dot{\nu} = \nu \circ \phi^{-1}$ . Then  $\dot{\nu}$  is stationary and ergodic. Now we apply Theorem 2 to this probability measure. Then we have

$$P_\omega \left\{ \frac{S_n}{n} \rightarrow \int \int x d\dot{F}(x) \pi(dF) \right\} = 1 \quad \nu - \text{a.e. } \omega,$$

Note that  $\int \int x d\dot{F}(x) \pi(dF) = \int \int \frac{x \exp(\xi x)}{M_F(\xi)} dF(x) \pi(dF) = f(\xi) = q$ . Hence the lemma follows.

Now back to the proof of Theorem 3. By the above lemma we have, for given  $\epsilon > 0$ , and  $\nu$ -a.e.  $\omega$  that for  $n \geq N(\omega)$

$$\frac{1}{n} \log P_\omega \left\{ \frac{S_n}{n} \in (q - \delta, q + \delta) \right\} \geq \frac{1}{n} \sum_{i=1}^n \log M_{F_i^\omega}(\xi) - \xi(q + \delta) - \epsilon$$

and consequently

$$\liminf_{n \rightarrow \infty} \frac{1}{n} \log P_\omega \left\{ \frac{S_n}{n} \in (q - \delta, q + \delta) \right\} \geq C(\xi) - \xi(q + \delta), \quad \nu - \text{a.e. } \omega.$$

By monotonicity, the result holds with  $\delta = 0$ , i.e., with  $K(q)$ .

We must now handle the general case. Suppose that there exists  $F \in \mathcal{F}$  such that  $F^{-1}$  is bounded. We replace all  $F$  by the distribution  $F * \phi_\epsilon$  where  $\phi_\epsilon(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x/\epsilon} \exp\left(-\frac{y^2}{2}\right) dy$  and apply the above results to this. Here  $*$  is the convolution. Then, letting  $\epsilon \downarrow 0$  the desired result follows.



REMARK 1. If  $\nu$  is i.i.d., then  $\{X_n\}$  is i.i.d. with respect to  $P$  with distribution function  $\overline{F}_1(x) = \int F(x)\pi(dF)$ . By the Cramer theorem,  $\{X_n\}$  with respect to  $P$  has the rate function

$$\overline{K}(x) = \sup\{\xi c - \overline{C}(\xi) | \xi \in R\}.$$

where  $\overline{C}(\xi) = \log \int \exp(\xi x) d\overline{F}(x) = \log \int \exp(\xi x) dF(x)\pi(dF)$ . By Jensen's inequality, we can check easily  $\overline{C}(\xi) \geq C(\xi)$  and hence  $\overline{K}(x) \leq K(x)$ .

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Department of Statistics  
 Taegu Hyosung Catholic University  
 Kyungbuk 713-702, Korea