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# Large deviations for processes on half-line<sup>\*</sup>

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#### Abstract

We consider a sequence of processes  $X_n(t)$  defined on the half-line  $0 \le t < \infty$ ,  $n = 1, 2, \ldots$  We give sufficient conditions for Large Deviation Principle (LDP) to hold in the space of continuous functions with metric

$$\rho_{\kappa}(f,g) = \sup_{t \ge 0} \frac{|f(t) - g(t)|}{1 + t^{1+\kappa}}, \quad \kappa \ge 0.$$

LDP is established for Random Walks and Diffusions defined on the half-line. LDP in this space is "more precise" than that with the usual metric of uniform convergence on compacts.

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### **1** Introduction

In this work we give sufficient conditions for a sequence of stochastic processes  $X_n(t)$ ;  $0 \le t < \infty$ , n = 1, 2, ..., to satisfy the Large Deviation Principle (LDP) in the space of continuous functions on  $[0, \infty)$ , which we denote by C. In the recent literature [18], [9], [12] the space  $\mathbb{C}$  is considered with the metric

$$\rho^{(P)}(f,g) := \sum_{k=1}^{\infty} 2^{-k} \min\{ \sup_{0 \le t \le k} |f(t) - g(t)|, 1 \}.$$
(1.1)

Theorem 2.6 of [19] gives sufficient conditions for  $X_n$  to satisfy LDP in the space  $(\mathbb{C}, \rho^{(P)})$ . As noted in [19], convergence  $f_n \to f$  in metric  $\rho^{(P)}$  is equivalent to convergence in  $\mathbb{C}[0,T]$  with uniform metric for any  $T \ge 0$ . A considerable drawback of metric  $\rho^{(P)}$  is that it is " not sensitive" to behaviour of functions as  $t \to \infty$ .

We consider the space  $\mathbb{C}$  with metric

$$\rho(f,g) = \rho_{\kappa}(f,g) := \sup_{t \ge 0} \frac{|f(t) - g(t)|}{1 + t^{1+\kappa}},$$

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for a fixed  $\kappa \ge 0$ . It is obvious that  $(\mathbb{C}, \rho)$  is a complete separable metric (Polish) space. As we shall see in §2, the LDP in the space  $(\mathbb{C}, \rho)$  is "more precise" than the LDP in  $(\mathbb{C}, \rho^{(P)})$ .

Here we treat continuous processes on infinite interval, a treatment of discontinuous processes on infinite interval will need a metric essentially different to  $\rho$ , (see [11], for the LDP for Compound Poisson processes on infinite interval). Note that in [10], Theorem 1.3.27, LDP for Wiener process in space ( $\mathbb{C}, \rho_{\kappa}$ ) when  $\kappa = 0$  is given, while in [7] the Law of Iterated Logarithm is proved for Wiener process in this space.

The paper is organised as follows. Sufficient conditions for LDP in the space  $(\mathbb{C}, \rho)$  are given in §2, Theorem 2.1. We also compare Theorem 2.1 and Theorem 2.6 of [19], and show that Theorem 2.1 is more precise. We apply Theorem 2.1 to Random Walks and Diffusions on the half line. Only Random Walks case is given here, and the reader is referred to the Arxiv for other examples.

#### 2 Main Result

To formulate the main result we need the following definitions and notations. For any  $T \in (0, \infty)$  denote by  $\mathbb{C}[0, T]$  the metric space of real continuous functions f = f(t);  $0 \le t \le T$ , with metric

$$\rho_T(f,g) := \sup_{0 \le t \le T} \frac{|f(t) - g(t)|}{1 + t^{1+\kappa}},$$

where  $\kappa \geq 0$  is fixed. We say that

$$I_0^T = I_0^T(f) : \mathbb{C}[0,T] \to [0,\infty],$$

is a (good) rate function in space  $\mathbb{C}[0,T]$  if: (*i*) it is lower semi-continuous: for any  $f \in \mathbb{C}[0,T]$ 

$$\lim_{f_n \to f} I_0^T(f_n) \ge I_0^T(f); \tag{2.1}$$

(*ii*) for any  $r \ge 0$  the set

$$B_{T,r} := \{ f \in \mathbb{C}[0,T] : I_0^T(f) \le r \}$$

is a compact in  $\mathbb{C}[0,T]$ .

For a non-empty measurable set  $B \subset \mathbb{C}[0,T]$  let

$$I_0^T(B) := \inf_{f \in B} I_0^T(f), \text{ with } I_0^T(\emptyset) := \infty.$$

 $(f)_{T,\varepsilon}$  and  $(B)_{T,\varepsilon}$  denote  $\varepsilon$ -neighbourhood in metric  $\rho_T$  in space  $\mathbb{C}[0,T]$  of  $f \in \mathbb{C}[0,T]$  and a measurable set  $B \subset \mathbb{C}[0,T]$  respectively. The interior and the closure of B is denoted by  $(B)_T$  and  $[B]_T$  respectively.

Note that lower semi-continuity (2.1) can be written as: for any  $f \in \mathbb{C}[0,T]$ 

$$\lim_{\varepsilon \to 0} I_0^T((f)_{T,\varepsilon}) = I_0^T(f).$$
(2.2)

For a function  $f \in \mathbb{C}$ ,  $f^{(T)}$  denotes its projection on  $\mathbb{C}[0, T]$ ,

$$f^{(T)} = f^{(T)}(t) := f(t); \ 0 \le t \le T$$

Denote by  $\mathbb{C}_0 \subset \mathbb{C}$  – the class of functions  $f \in \mathbb{C}$ , such that f(0) = 0,  $\lim_{t \to \infty} \frac{f(t)}{1+t^{1+\kappa}} = 0$ .

Let now  $X_n(t)$ ;  $t \in [0, \infty)$ , n = 1, 2, ..., be a sequence of processes in space  $\mathbb{C}_0$ . We assume the following conditions.

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I. For any  $T \in (0,\infty)$  processes  $X_n^{(T)}$  satisfy LDP in space  $\mathbb{C}[0,T]$  with good rate function  $I_0^T$ , i.e. for any measurable set  $B \subset \mathbb{C}[0,T]$ 

$$\overline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n^{(T)} \in B) \leq -I_0^T([B]_T),$$

$$\underline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n^{(T)} \in B) \geq -I_0^T((B)_T).$$

Moreover, for any  $f \in \mathbb{C}[0,T]$  there is  $g = g_f \in \mathbb{C}_0$ , such that  $g^{(T)} = f$ , and for any  $U \ge T$  it holds

$$I_0^U(g^{(U)}) = I_0^T(f).$$
(2.3)

Condition (2.3) means that one can extend any  $f \in \mathbb{C}[0,T]$  for t > T such that the rate function will stay the same. It is natural to call the function  $g = g_f$  the most likely extension of f beyond [0,T].

**II**. For any  $r \ge 0$ 

$$\lim_{T\to\infty}\sup_{f\in B^+_r}\sup_{t\geq T}\frac{|f(t)|}{1+t^{1+\kappa}}=0,$$

where

$$B_r^+ := \{ f \in \mathbb{C} : \lim_{T \to \infty} I_0^T(f^{(T)}) \le r \}.$$

III. For any  $N < \infty$  and  $\varepsilon > 0$  there is  $T = T_{N,\varepsilon} < \infty$  such that

$$\overline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(\sup_{t \ge T} \frac{|X_n(t)|}{1 + t^{1+\kappa}} > \varepsilon) \le -N.$$

**Theorem 2.1.** Assume Conditions I, II and III. Then for any  $f \in \mathbb{C}$  there exists

$$\lim_{T \to \infty} I_0^T(f^{(T)}) =: I(f),$$
(2.4)

and it is a good rate function in the space  $(\mathbb{C}, \rho)$ . The sequence  $X_n$  satisfies LDP in this space with rate function I(f), i.e. for any measurable  $B \subset \mathbb{C}$ 

$$\overline{\lim_{n \to \infty}} \, \frac{1}{n} \ln \mathbf{P}(X_n \in B) \le -I([B]), \tag{2.5}$$

$$\underline{\lim}_{n \to \infty} \frac{1}{n} \ln \mathbf{P}(X_n \in B) \ge -I((B)), \tag{2.6}$$

where

$$I(B) := \inf_{f \in B} I(f), \text{ with } I(\emptyset) = \infty.$$

Before we give the proof, we compare our Theorem to Theorem 2.6 in [19], which gives sufficient conditions for LDP in the space  $(\mathbb{C}, \rho^{(P)})$ . Note that if a set B is such that I([B]) = I((B))(=I(B)), then inequalities (2.5), (2.6) can be replaced by the equality

$$\lim_{n \to \infty} \frac{1}{n} \ln \mathbf{P}(X_n \in B) = -I(B).$$

Hence the difference

$$D(B) := I((B)) - I([B]) \ge 0$$

describes the precision of LDP: the smaller the difference the more precise is the theorem.

The rate functions in both theorems are the same. This is because projections  $X_n^{(T)}$  on [0,T] satisfy LDP in the space  $\mathbb{C}[0,T]$  with uniform metric and common rate function  $I_0^T(f)$ . Therefore we can compare these theorems by comparing differences

$$D(B) := I((B)) - I([B])$$
 and  $D^{(P)}(B) := I((B)^{(P)}) - I([B]^{(P)})$ 

where  $[B]^{(P)}$ ,  $(B)^{(P)}$  denote the closure and the interior of B in metric  $\rho^{(P)}$ .

As noted earlier, convergence in  $\rho^{(P)}$  is equivalent to convergence in  $\rho_T$  for any T > 0. Therefore  $\rho(f_n, f) \to 0$  implies  $\rho^{(P)}(f_n, f) \to 0$ . It is easy to see that the opposite is not true.

Thus

$$[B] \subset [B]^{(P)}, \quad (B)^{(P)} \subset (B),$$

and therefore

$$I([B]^{(P)}) \le I([B]), \quad I((B)) \le I((B)^{(P)}),$$

so that we always have  $D(B) \leq D^{(P)}(B)$ . Below we give an example of B satisfying simultaneously

$$I([B]) = I((B)) \in (0, \infty), \quad I([B]^{(P)}) = 0.$$

Hence Theorem 2.1 allows to give "precise" logarithmic asymptotic for  $\mathbf{P}(X_n \in B)$ , while Theorem 2.6 in [19] does not. Thus we conclude that *LDP* in the space  $(\mathbb{C}, \rho)$  is more precise than in the space  $(\mathbb{C}, \rho^{(P)})$ .

**Example 2.2.** Consider Wiener process w = w(t) on  $[0, \infty)$ . Denote

$$w_n = w_n(t) := \frac{1}{\sqrt{n}}w(t), \quad t \ge 0.$$

Since conditions I—III are easily checked, then LDP follows from Theorem 2.1 with rate function

 $I(f) = \left\{ \begin{array}{ll} \frac{1}{2} \int_0^\infty (f'(t))^2 dt, & \text{if} \quad f(0) = 0, \quad f \text{ is absolutely continuous}, \\ \infty & \text{otherwise}. \end{array} \right.$ 

$$B = \overline{(f_0)_1} := \left\{ g \in \mathbb{C} : \sup_{t \ge 0} \frac{|g(t)|}{1+t} \ge 1 \right\}, \quad f_0 = f_0(t) \equiv 0$$

Since it is a complement to an open set  $(f_0)_1$ , it is closed in  $(\mathbb{C}, \rho)$ , and therefore

$$I([B]) = I(B) = \inf_{g \in B} I(g).$$

By Cauchy-Schwarz-Bunyakovski inequality

$$1 \leq \sup_{t \geq 0} \frac{|g(t)|}{1+t} = \sup_{t \geq 0} \frac{|\int_0^t g'(s)ds|}{1+t} \leq \sup_{t \geq 0} \left| \int_0^t (g'(s))^2 ds \right|^{1/2} \sup_{t \geq 0} \frac{t^{1/2}}{1+t} = \frac{\sqrt{2I(g)}}{2}.$$

Hence  $I(g) \ge 2$  for all  $g \in B$ .

Take now  $f(t) = 2tI(0 \le t \le 1) + 2I(t \ge 1)$ . It is easy to see that  $f \in B$  and I(f) = 2. Therefore I([B]) = I(B) = 2.

Taking  $f_n(t) = (2 + 1/n)tI(0 \le t \le 1) + 2 + 1/nI(t \ge 1)$ , we can see that  $f_n \in (B)$  and I([B]) = I(B) = I((B)) = 2. Hence,

$$\lim_{n \to \infty} \frac{1}{n} \ln \mathbf{P}(w_n \in B) = -2.$$

Consider now  $[B]^{(P)}$ , the closure of B in metric  $\rho^{(P)}$ . By taking  $g_n(t) = \frac{t^2}{n}$  it is easy to see that  $g_n \in B$  for all n and  $\lim_{n \to \infty} \rho^{(P)}(g_n, f_0) = 0$ . Therefore,  $f_0 \in [B]^{(P)}$ , and  $I([B]^{(P)}) = 0$ . Hence the upper bound for this set is trivial, and does not allow to find logarithmic asymptotic of the required probability.

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# **3 Proof of Theorem 2.1**

Denote by  $(f)_{\varepsilon}$  and  $(B)_{\varepsilon}$  the  $\varepsilon$ -neighborhoods ( $\varepsilon > 0$ ) of  $f \in \mathbb{C}$ , and set  $B \subset \mathbb{C}$ . The proof of the Theorem 2.1 consists of three steps. The first step in Lemma 3.1 proves that I(f) is a good rate function. The second step proves the local LDP for  $X_n$  in  $(\mathbb{C}_0, \rho)$  in Lemma 3.2 and Corollary 3.3. The third step proves a weaker form of exponential tightness for  $X_n$  in Lemma 3.4. The proof is then concluded as follows.

The upper bound is obtained by Lemmas 3.2 and 3.4 from general results on LDP in metric spaces: for  $B \subset \mathbb{C}_0$  and  $\varepsilon > 0$  it holds (e.g. [4], Theorem 3.1)

$$\overline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n \in B) \le -I((B)_{\varepsilon}).$$

It is also known (e.g. [4], Lemma 2.1), that a good rate function I(f) satisfies

$$\lim_{\varepsilon \to 0} I((B)_{\varepsilon}) = I([B]),$$

and the upper bound (2.5) follows. Lower bound (2.6) follows from Lemma 3.2.

**Lemma 3.1.** The rate function I(f) (defined in (2.4)) is a good rate function, i.e. it is lower semi-continuous and for any  $r \ge 0$  the set  $B_r := \{f \in \mathbb{C} : I(f) \le r\}$  is a compact in  $\mathbb{C}$ .

*Proof.* First we show that the limit exists. It is known (see e.g. [4], Theorem 3.1 or Lemma 1.3), that LDP implies local LDP: for any  $f \in \mathbb{C}[0,T]$ 

$$-I_0^T(f) \ge \lim_{\varepsilon \to 0} \lim_{n \to \infty} \frac{1}{n} \ln \mathbf{P}(X_n^{(T)} \in (f)_{T,\varepsilon}) \ge \lim_{\varepsilon \to 0} \lim_{n \to \infty} \frac{1}{n} \ln \mathbf{P}(X_n^{(T)} \in (f)_{T,\varepsilon}) \ge -I_0^T(f).$$

For  $U \geq T$ , with obvious notations, we have for  $f \in \mathbb{C}$ 

$$\{X_n^{(U)} \in (f^{(U)})_{U,\varepsilon}\} \subset \{X_n^{(T)} \in (f^{(T)})_{T,\varepsilon}\},\$$

therefore

$$-I_0^T(f^{(T)}) \geq \lim_{\varepsilon \to 0} \overline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n^{(T)} \in (f^{(T)})_{T,\varepsilon})$$
  
$$\geq \lim_{\varepsilon \to 0} \lim_{n \to \infty} \frac{1}{n} \ln \mathbf{P}(X_n^{(U)} \in (f^{(U)})_{U,\varepsilon}) \geq -I_0^U(f^{(U)})$$

Thus we established that  $I_0^T(f^{(T)})$  is non-decreasing in T, and (2.4) follows. Next, we show lower semi-continuity, that is if  $f_n \to f$ , then

$$\underbrace{\lim_{n \to \infty} I(f_n) \ge I(f).}$$
(3.1)

For any  $N < \infty$ ,  $\varepsilon > 0$  there is  $T = T_{N,\varepsilon} < \infty$  such that

$$I_0^T(f) \ge \min\{I(f), N\} - \varepsilon.$$

Since  $f_n \to f$ ,  $\rho_T(f_n, f) \to 0$ . Thanks to Condition I the rate function  $I_0^T(f)$  is lower semi-continuous in  $(\mathbb{C}[0, T], \rho_T)$ , therefore

$$\lim_{n \to \infty} I(f_n) \ge \lim_{n \to \infty} I_0^T(f_n) \ge I_0^T(f) \ge \min\{I(f), N\} - \varepsilon.$$

This implies (3.1), since N and  $\varepsilon$  are arbitrary.

We show next that the set  $B_r$  is completely bounded. Due to Condition II for any  $\varepsilon > 0$  there is  $T = T_r < \infty$  such that for any  $f \in B_r$ 

$$\sup_{t \ge T} \frac{|f(t)|}{1 + t^{1+\kappa}} < \varepsilon.$$
(3.2)

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Denote

$$B_r^{(T)} := \{ f^{(T)} : f \in B_r \},\$$

so that

$$B_r^{(T)} \subset B_{T,r},$$

where we recall that  $B_{T,r} := \{ f \in \mathbb{C}[0,T] : I_0^T(f) \le r \}.$ 

Since by Condition I the set  $B_{T,r}$  is a compact in  $\mathbb{C}[0,T]$ , it is possible to find finite  $\varepsilon$ -net:

$$B_r^{(T)} \subset B_{T,r} \subset \bigcup_{i=1}^M (f_i)_{T,\varepsilon}$$

Now for  $f \in \mathbb{C}[0,T]$  define  $f^{(T+)} \in \mathbb{C}_0$  as

$$f^{(T+)}(t) := \begin{cases} f(t), & \text{if } 0 \le t \le T, \\ f(T), & \text{if } t \ge T \end{cases}$$

For any  $f \in B_r$  there is  $i \in \{1, \cdots, M\}$  such that

$$\sup_{0 \le t \le T} \frac{|f(t) - f_i^{(T+)}(t)|}{1 + t^{1+\kappa}} < \varepsilon < 3\varepsilon.$$

We have for this i due to (3.2)

$$\sup_{t \ge T} \frac{|f(t) - f_i^{(T+)}(t)|}{1 + t^{1+\kappa}} \le \sup_{t \ge T} \frac{|f(t)|}{1 + t^{1+\kappa}} + \sup_{t \ge T} \frac{|f(T)|}{1 + t^{1+\kappa}} + \sup_{t \ge T} \frac{|f(T) - f_i^{(T+)}(T)|}{1 + t^{1+\kappa}} \le 3\varepsilon,$$

therefore the collection  $\{f_1^{(T+)}, \cdots, f_M^{(T+)}\}$  represents a  $3\varepsilon$ -net in the set  $B_r$ . Thus we have shown that the set  $B_r$  is completely bounded in  $\mathbb{C}_0$ .

From lower semi-continuity of I(f), established earlier, it follows that  $B_r$  is closed in  $\mathbb{C}_0$ . Since a closed completely bounded subset of a Polish space is a compact (e.g. [15], Theorem 3, p. 109), we have shown that  $B_r$  is a compact in  $\mathbb{C}_0$ , thus completing the proof of Lemma 3.1.

**Lemma 3.2.** For any  $f \in \mathbb{C}_0$ ,  $\varepsilon > 0$ 

$$\overline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n \in (f)_{\varepsilon}) \le -I((f)_{2\varepsilon}), \tag{3.3}$$

$$\underline{\lim}_{n \to \infty} \frac{1}{n} \ln \mathbf{P}(X_n \in (f)_{\varepsilon}) \ge -I(f).$$
(3.4)

*Proof.* First we prove the lower bound (3.4) as it is also used in the proof of the upper bound. If  $I(f) = \infty$ , then (3.4) is trivially satisfied. Let now  $I(f) < \infty$ . For any  $T \in (0, \infty)$  we have

$$\{X_n \in (f)_{\varepsilon}\} \supset A_n(T) \cap B_n(T) \cap C_n(T) \cap D(T) = A_n(T) \cap B_n(T) \cap D(T),$$

where

$$\begin{split} A_n(T) &:= \left\{ \sup_{0 \le t \le T} \frac{|X_n(t) - f(t)|}{1 + t^{1+\kappa}} < \varepsilon \right\}, \quad B_n(T) &:= \left\{ \sup_{t \ge T} \frac{|X_n(t) - X_n(T)|}{1 + t^{1+\kappa}} < \varepsilon/4 \right\}, \\ C_n(T) &:= \left\{ \frac{|X_n(T)|}{1 + T^{1+\kappa}} < \varepsilon/4 \right\}, \quad D(T) &:= \left\{ \sup_{t \ge T} \frac{|f(t)|}{1 + t^{1+\kappa}} < \varepsilon/2 \right\}. \end{split}$$

For a large T the event D(T) is a certainty (due to  $I(f)<\infty$ ). Therefore there exists  $T_0<\infty$ , such that for all  $T\geq T_0$ 

$$\mathbf{P}(X_n \in (f)_{\varepsilon}) \ge \mathbf{P}(A_n(T) \cap B_n(T)) \ge \mathbf{P}(A_n(T)) - \mathbf{P}(\overline{B_n}(T)),$$
(3.5)

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where  $\overline{B_n}(T)$  is a complement of  $B_n(T)$ . Due to Condition III there is  $T \ge T_0$  such that

$$\overline{\lim_{n \to \infty}} \, \frac{1}{n} \ln \mathbf{P}(\overline{B_n}(T)) \le -2I(f), \tag{3.6}$$

and for this T due to Condition I we have

$$\underbrace{\lim_{n \to \infty} \frac{1}{n} \ln \mathbf{P}(A_n(T))}_{n \to \infty} \ge -I_0^T(f^{(T)}) \ge -I(f).$$
(3.7)

(3.4) now follows from (3.5) by using (3.6), (3.7).

Next we prove the upper bound (3.3). It is obvious that for any  $T \in (0, \infty)$ 

$$\mathbf{P}(X_n \in (f)_{\varepsilon}) \le \mathbf{P}(X_n^{(T)} \in (f^{(T)})_{T,\varepsilon}),$$

where we recall that  $(f)_{T,\varepsilon}$  denote  $\varepsilon$ -neighbourhood in metric  $\rho_T$  in space  $\mathbb{C}[0,T]$  of  $f \in \mathbb{C}[0,T].$ 

Due to Condition I for any  $\delta > 0$ 

$$L(\varepsilon) := \lim_{n \to \infty} \frac{1}{n} \ln \mathbf{P}(X_n \in (f)_{\varepsilon}) \leq \lim_{n \to \infty} \frac{1}{n} \ln \mathbf{P}(X_n^{(T)} \in (f^{(T)})_{T,\varepsilon}) \\ \leq -I_0^T([(f^{(T)})_{T,\varepsilon}]_T) \leq -I_0^T((f^{(T)})_{T,\varepsilon+\delta}).$$

For any  $T \in (0,\infty)$  and chosen  $\varepsilon$  and  $\delta$ , in this way we have the inequality

$$L(\varepsilon) \le -I_0^T((f^{(T)})_{T,\varepsilon+\delta}).$$
(3.8)

Choose now  $T < \infty$  so large, that simultaneously the following holds:

$$\sup_{t\ge T}\frac{|f(t)|}{1+t^{1+\kappa}}<\delta;$$
(3.9)

$$\overline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n \in R(T, \varepsilon)) \le -N,$$
(3.10)

where  $N < \infty$  is arbitrary, and

$$R(T,\varepsilon) := \left\{ g \in \mathbb{C}_0 : \sup_{t \ge T} \frac{|g(t)|}{1 + t^{1+\kappa}} > \varepsilon \right\}.$$

Denote

$$((f^{(T)})_{T,\varepsilon+\delta})^{(T+)} := \{g \in \mathbb{C}_0 : g^{(T)} \in (f^{(T)})_{T,\varepsilon+\delta}\}$$

Next we show that

$$I_0^T(B) = I(B^{(T+)}), (3.11)$$

where for  $B \subset \mathbb{C}[0,T]$ 

$$B^{(T+)} := \{ g \in \mathbb{C}_0 : g^{(T)} \in B \}.$$

Indeed, for any  $\varepsilon > 0$  let  $f \in B$  be such that

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$$I_0^T(f) \le I_0^T(B) + \varepsilon$$

Then due to (2.3) in condition I there is  $g \in \mathbb{C}_0$  such that  $g^{(T)} = f$  (consequently  $g \in B^{(T+)}$ ) with  $I(g) = I_0^T(f)$ . Therefore

$$I_0^T(B) + \varepsilon \ge I_0^T(f) = I(g) \ge I(B^{(T+)}).$$

Since  $\varepsilon > 0$  is arbitrary,

$$I_0^T(B) \ge I(B^{(T+)}).$$
 (3.12)

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Let now  $g\in B^{(T+)}$  such that

$$I(g) \leq I(B^{(T+)}) + \varepsilon$$

Then  $g^{(T)} \in B$  with  $I_0^T(g^{(T)}) \leq I(g)$ . Therefore

$$I(B^{(T+)}) + \varepsilon \ge I(g) \ge I_0^T(g^{(T)}) \ge I_0^T(B),$$

and

$$I_0^T(B) \le I(B^{(T+)}). \tag{3.13}$$

Inequalities (3.12), (3.13) now prove equality (3.11).

Due to (3.11) we have

$$I_0^T((f^{(T)})_{T,\varepsilon+\delta}) = I(((f^{(T)})_{T,\varepsilon+\delta})^{(T+)}),$$

therefore due to (3.8)

$$L(\varepsilon) \le -I(((f^{(T)})_{T,\varepsilon+\delta})^{(T+)}).$$
(3.14)

Take an arbitrary  $g \in ((f^{(T)})_{T,\varepsilon+\delta})^{(T+)}$ . Then either

$$\sup_{t\geq T}\frac{|g(t)-f(t)|}{1+t^{1+\kappa}}<\varepsilon+2\delta,$$

and then

$$g \in (f)_{\varepsilon + 2\delta};\tag{3.15}$$

or

$$\sup_{t>T} \frac{|g(t) - f(t)|}{1 + t^{1+\kappa}} \ge \varepsilon + 2\delta, \tag{3.16}$$

and then

$$\sup_{t \ge T} \frac{|g(t)|}{1 + t^{1+\kappa}} \ge \varepsilon + \delta, \tag{3.17}$$

and

$$g \in R(T,\varepsilon). \tag{3.18}$$

To see how (3.17) follows from (3.16), note that if the inequality (3.17) is not true, then the opposite holds

$$\sup_{t \ge T} \frac{|g(t)|}{1 + t^{1 + \kappa}} < \varepsilon + \delta,$$

and due to (3.9)

$$\sup_{t\geq T}\frac{|g(t)-f(t)|}{1+t^{1+\kappa}}\leq \sup_{t\geq T}\frac{|g(t)|}{1+t^{1+\kappa}}+\sup_{t\geq T}\frac{|f(t)|}{1+t^{1+\kappa}}<\varepsilon+\delta+\delta=\varepsilon+2\delta,$$

which contradicts (3.16). We have proved (see (3.15) and (3.18)), that

 $((f^{(T)})_{T,\varepsilon+\delta})^{(T+)} \subset (f)_{\varepsilon+2\delta} \cup R(T,\varepsilon).$ 

From the latter we obtain

$$I(((f^{(T)})_{T,\varepsilon+\delta})^{(T+)}) \ge \min\{I((f)_{\varepsilon+2\delta}), I(R(T,\varepsilon))\}.$$
(3.19)

Further, due to (3.10)

$$-N \ge \overline{\lim_{n \to \infty} \frac{1}{n}} \ln \mathbf{P}(X_n \in R(T, \varepsilon)) \ge \underline{\lim_{n \to \infty} \frac{1}{n}} \ln \mathbf{P}(X_n \in R(T, \varepsilon)) \ge -I(R(T, \varepsilon)),$$

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where the last inequality for an open set  $R(T, \varepsilon)$  follows from the established lower bound (3.4). Therefore  $I(R(T, \varepsilon)) \ge N$ , and, in view of (3.19),

$$I(((f^{(T)})_{T,\varepsilon+\delta})^{(T+)}) \ge \min\{I((f)_{\varepsilon+2\delta}), N\}.$$

Going back to (3.14), we obtain the inequality

$$L(\varepsilon) \le -\min\{I((f)_{\varepsilon+2\delta}), N\},\$$

in which  $\delta > 0$  and  $N < \infty$  are arbitrary. Taking  $2\delta = \varepsilon$  and sending N to  $\infty$ , we obtain the required upper bound

$$L(\varepsilon) \le -I((f)_{2\varepsilon}).$$

Local LDP in  $\mathbb{C}_0$  follows from Lemma 3.2, and is stated as a corollary. Corollary 3.3. For any  $f \in \mathbb{C}_0$ 

$$\lim_{\varepsilon \to 0} \overline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n \in (f)_{\varepsilon}) \le -I(f), \quad \lim_{\varepsilon \to 0} \underline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n \in (f)_{\varepsilon}) \ge -I(f).$$

Next result proves a weaker form of exponential tightness: for any N there is a completely bounded set  $K_N$  in  $(\mathbb{C}_0, \rho)$  such that

$$\overline{\lim_{n \to \infty} \frac{1}{n}} \ln \mathbf{P}(X_n \notin K) \le -N.$$

**Lemma 3.4.** For any  $N < \infty$  and  $\varepsilon > 0$  there is a finite collection of  $g_1, \dots, g_M \in \mathbb{C}_0$  such that

$$\overline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n \notin \bigcup_{i=1}^M (g_i)_{\varepsilon}) \le -N.$$
(3.20)

Proof. Denote by

$$R_T(\varepsilon) := \left\{ f \in \mathbb{C}_0 : \sup_{t \ge T} \frac{|f(t)|}{1 + t^{1+\kappa}} \le \varepsilon \right\}.$$

Then due to condition III there is  $T < \infty$  such that

$$\overline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n \notin R_T(\varepsilon)) \le -N.$$
(3.21)

For this T, thanks to condition I, processes  $X_n^{(T)}$  satisfy LDP in the space  $\mathbb{C}[0, T]$ . Therefore for a chosen N by the Puhalskii theorem on exponential tightness ([19], Theorem 2.1) there is a compact  $K \subset \mathbb{C}[0, T]$  such that

$$\overline{\lim_{n \to \infty}} \, \frac{1}{n} \ln \mathbf{P}(X_n^{(T)} \notin K) \le -N.$$

For a given  $\varepsilon > 0$  take a finite  $\varepsilon$ -net  $f_1, \dots, f_M \in \mathbb{C}[0, T]$  in K:

$$K \subset \cup_{i=1}^{M} (f_i)_{T,\varepsilon}.$$

Then

$$\overline{\lim_{n \to \infty}} \frac{1}{n} \ln \mathbf{P}(X_n^{(T)} \notin \bigcup_{i=1}^M (f_i)_{T,\varepsilon}) \le -N.$$
(3.22)

Denote for all  $i = 1, \cdots, M$ 

$$g_i(t) := \begin{cases} f_i(t), & \text{if } 0 \le t \le T, \\ f_i(T), & \text{if } t \ge T. \end{cases}$$

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Define the set  $\mathcal{M}_{\varepsilon} := \{i \in \{1, \cdots, M\} : \frac{|f_i(T)|}{1+T^{1+\kappa}} \leq 2\varepsilon\}$ . Then

$$P := \mathbf{P}(X_n \notin \bigcup_{i=1}^M (g_i)_{3\varepsilon}) \le \mathbf{P}(X_n \notin R_T(\varepsilon)) + \mathbf{P}(X_n \in R_T(\varepsilon), \ X_n^{(T)} \notin \bigcup_{i=1}^M (f_i)_{T,\varepsilon}) + \mathbf{P}(X_n \in R_T(\varepsilon), \ X_n^{(T)} \in \bigcup_{i=1}^M (f_i)_{T,\varepsilon}, \ X_n \notin \bigcup_{i=1}^M (g_i)_{3\varepsilon}) =: P_1 + P_2 + P_3.$$

We bound  $P_3$  as follows:

$$P_3 \leq \sum_{i \in \mathcal{M}_{\varepsilon}} \mathbf{P}\left(\sup_{t \geq T} \frac{|X_n(t) - f_i(T)|}{1 + t^{1+\kappa}} > 3\varepsilon\right) \leq M \mathbf{P}\left(\sup_{t \geq T} \frac{|X_n(t)|}{1 + t^{1+\kappa}} > \varepsilon\right) \leq M \mathbf{P}(X_n \notin R_T(\varepsilon)).$$

Since

$$P_1 \leq \mathbf{P}(X_n \notin R_T(\varepsilon)),$$

we obtain

$$P \le \mathbf{P}(X_n^{(T)} \notin \bigcup_{i=1}^M (f_i)_{T,\varepsilon}) + (M+1)\mathbf{P}(X_n \notin R_T(\varepsilon)).$$
(3.23)

The required inequality in (3.20) now follows from (3.21), (3.22) and (3.23).  $\hfill \Box$ 

## 4 Large Deviations for Random Walks

#### 4.1 Large Deviation Principle for Random Walks on half-line.

Let  $\xi$  be a non-degenerate random variable satisfying the following condition  $[\mathbf{C}_{\infty}]$ . For any  $\lambda \in \mathbb{R}$   $\mathbf{E}e^{\lambda\xi} < \infty$ .

Denote by  $A(\lambda) := \ln \mathbf{E} e^{\lambda \xi}$  the log moment generating function, and by  $\Lambda(\alpha)$  the Deviation function of  $\xi$  (Legendre-Fenchel transform of A)

$$\Lambda(\alpha) := \sup_{\lambda} \{\lambda \alpha - A(\lambda)\}.$$

Denote  $S_0 := 0$ ,  $S_k := \xi_1 + \cdots + \xi_k$  for  $k \ge 1$ , where  $\{\xi_n\}$  is a sequence of i.i.d. copies of  $\xi$ . Consider a random piece-wise linear function  $s_n = s_n(t) \in \mathbb{C}$ , going through the nodes

$$\left(\frac{k}{n}, \frac{S_k}{x}\right), \quad k = 0, 1, \cdots,$$

where x = x(n) is a sequence of positive constants such that  $x \sim n$ ,  $n \to \infty$ .

**Theorem 4.1.** Assume  $[\mathbf{C}_{\infty}]$ . Then the sequence of processes  $s_n(t)$  satisfies LDP in space  $(\mathbb{C}, \rho_{\kappa})$  with  $\kappa = 0$  and rate function I,

$$I(f) := \left\{ \begin{array}{ll} \int_0^\infty \Lambda(f'(t)) dt, \quad f(0) = 0, \quad f \ \text{ is absolutely continuous}, \\ +\infty, \text{ otherwise}. \end{array} \right.$$

Proof. Without loss of generality we can take  $\mathbf{E}\xi = 0$ . This is because the Deviation function for  $\xi^{(0)} := \xi - a$  is given by  $\Lambda^{(0)}(\alpha) = \Lambda(\alpha + a)$ . (superscript <sup>(0)</sup> denotes quantities for the centered random variable). Therefore the rate function for  $s_n^{(0)}$ , is given by  $I^{(0)}(f) = I(f + e_a)$ , where  $e_a = e_a(t) := at$ ;  $t \ge 0$ . Clearly,  $s_n = s_n^{(0)} + e_a$ ,  $\mathbf{P}(s_n \in B) = \mathbf{P}(s_n^{(0)} \in B - e_a)$ , where  $B - e_a := \{f - e_a : f \in B\}$ . It is obvious that  $[B - e_a] = [B] - e_a$ ,  $(B - e_a) = (B) - e_a$ , implying  $I^{(0)}([B - e_a]) = I([B])$ ,  $I^{(0)}((B - e_a)) = I((B))$ . Hence LDP for  $s_n^{(0)}$  with rate function  $I^{(0)}$  implies LDP for  $s_n$  with rate function I.

The rest of the proof consists in checking Conditions I - III. Condition I follows from the LDP for  $s_n$  in  $\mathbb{C}[0,1]$  (Theorem 9 of [5] or [4], Section 6.2).

Proof of II. By  $[\mathbf{C}_{\infty}]$ , with  $\mathbf{E}\xi = 0$  it follows that there exists a non-decreasing continuous function h(t);  $t \ge 0$ , such that for some  $\delta > 0$ ,  $h(t) = \delta t$ , if  $0 \le t \le 1$ ,  $\lim_{t\to\infty} h(t) = \infty$ , and that for all  $\alpha \in \mathbb{R}$  the following inequality holds

$$\Lambda(\alpha) \ge h(|\alpha|)|\alpha|. \tag{4.1}$$

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Indeed, for  $\alpha \to 0$  (see e.g. [5], p.21)  $\Lambda(\alpha) \sim \frac{\alpha^2}{2\sigma^2}$ ; and for any  $\lambda > 0$ ,  $\alpha > 0$ 

 $\Lambda(\alpha) \geq \lambda \alpha - A(\lambda), \quad \Lambda(-\alpha) \geq \lambda \alpha - A(-\lambda).$ Therefore  $\lim_{|\alpha| \to \infty} \frac{\Lambda(\alpha)}{|\alpha|} \geq \lambda$ , so that  $\lim_{|\alpha| \to \infty} \frac{\Lambda(\alpha)}{|\alpha|} = \infty$ , and (4.1) follows. Denote by

$$f_T(t) := t \frac{f(T)}{T}, \quad t \in [0, T]$$

The function  $f_T$  "straightens" function f on [0, T]:

$$I_0^T(f) \ge I_0^T(f_T) = \int_0^T \Lambda\left(\frac{f(T)}{T}\right) dt = T\Lambda\left(\frac{f(T)}{T}\right).$$

Therefore by (4.1) for  $f \in B_r$ 

$$r \ge T\Lambda\left(\frac{f(T)}{T}\right) \ge T\frac{|f(T)|}{T}h\left(\frac{|f(T)|}{T}\right),$$
$$\frac{|f(T)|}{T} \le \frac{r}{Th\left(\frac{|f(T)|}{T}\right)}.$$

so that

Let 
$$c := \sqrt{\frac{T}{\delta}}$$
, and  $T$  be such that  $\frac{c}{\sqrt{T}} \leq 1$ . Assume that  $\frac{|f(T)|}{T} > \frac{c}{\sqrt{T}}$ . Then it follows from (4.2)

$$\frac{|f(T)|}{T} \leq \frac{r}{Th(\frac{c}{\sqrt{T}})} = \frac{r}{T\delta\frac{c}{\sqrt{T}}} = \frac{c}{\sqrt{T}},$$

which is a contradiction. Thus for  $T\geq c^2=\frac{r}{\delta}$  it holds

$$|f(T)| \le \sqrt{\frac{r}{\delta}}\sqrt{T}$$

Clearly,  $B_r = B_r^+$ . Therefore we have proved

$$\sup_{f \in B_r^+} \sup_{t \ge T} \frac{|f(t)|}{1+t} \le \sqrt{\frac{r}{\delta}} \frac{\sqrt{T}}{1+T} \le \sqrt{\frac{r}{\delta}} \frac{1}{\sqrt{T}}.$$

Condition **II** now follows.

Check now condition III. Let  $T_n := \max\{\frac{k}{n} \le T : k = 1, 2, \dots\}$ , to have

$$\begin{aligned} \mathbf{P}\left(\sup_{t\geq T} \frac{|s_n(t)|}{1+t} > \varepsilon\right) &\leq \mathbf{P}\left(\sup_{t\geq T_n} \frac{|s_n(t)|}{1+t} > \varepsilon\right) \\ &\leq \mathbf{P}\left(\sup_{t\geq T_n} \frac{|s_n(t) - s_n(T_n)|}{1+t} > \varepsilon/2\right) + \mathbf{P}\left(\sup_{t\geq T_n} \frac{|s_n(T_n)|}{1+t} > \varepsilon/2\right) \\ &= \mathbf{P}\left(\sup_{u\geq 0} \frac{|s_n(u)|}{1+T_n+u} > \varepsilon/2\right) + \mathbf{P}\left(\sup_{t\geq T_n} \frac{|s_n(T_n)|}{1+T_n} > \varepsilon/2\right) \\ &\leq \mathbf{P}\left(\sup_{k\geq 1} \frac{|s_n(\frac{k}{n})|}{T+\frac{k}{n}} > \varepsilon/2\right) + \mathbf{P}\left(\frac{|s_n(T_n)|}{T_n} > \varepsilon/2\right) =: P_1(n) + P_2(n).\end{aligned}$$

To bound  $P_1(n)$  use the exponential Chebyshev's (Chernoff's) inequality

$$P_{1}(n) \leq \sum_{k \geq 1} \mathbf{P}\left(\frac{|S_{k}|}{x(T+\frac{k}{n})} > \varepsilon/2\right)$$
  
$$\leq \sum_{k \geq 1} \mathbf{P}\left(\frac{S_{k}}{x(T+\frac{k}{n})} > \varepsilon/2\right) + \sum_{k \geq 1} \mathbf{P}\left(\frac{S_{k}}{x(T+\frac{k}{n})} < -\varepsilon/2\right)$$
  
$$\leq \sum_{k \geq 1} e^{-k\Lambda(R)} + \sum_{k \geq 1} e^{-k\Lambda(-R)},$$

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(4.2)

where  $R := \frac{x}{k}(T + \frac{k}{n})$ . Since for all *n* and *T* large enough

$$R \ge \varepsilon/4, \quad kR \ge T\varepsilon/4 + k\varepsilon/4,$$

we have due to (4.1) for  $\varepsilon/4 \in (0,1)$ 

$$k\Lambda(\pm R) \ge kRh(R) \ge (T\varepsilon/4 + k\varepsilon/4)\delta\varepsilon/4 = T\frac{\delta\varepsilon^2}{16} + k\frac{\delta\varepsilon^2}{16}.$$

Therefore

$$P_1(n) \le 2e^{-T\delta_1} \sum_{k\ge 1} e^{-k\delta_1} = C_1 e^{-T\delta_1},$$

where  $\delta_1 := \frac{\delta \varepsilon^2}{16}$ ,  $C_1 := 2 \frac{e^{-\delta_1}}{1 - e^{-s_1}}$ . Similarly we obtain the bound  $P_2(n) \leq C_2 e^{-T\delta_2}$  for some  $\delta_2 > 0$ ,  $C_2 < \infty$ . Condition III follows.

### 4.2 Moderate Deviation Principle for Random Walks on half-line.

Let  $s_n = s_n(\cdot) \in \mathbb{C}$  be defined as in previous section, and let  $\mathbf{E}\xi = 0$ . Assume Cramer's condition

[C<sub>0</sub>]. For some  $\delta > 0$   $\mathbf{E}e^{\delta|\xi|} < \infty$ .

Let a sequence x = x(n), used in the construction of  $s_n$ , satisfy as  $n \to \infty$ 

$$\frac{x}{\sqrt{n}} \to \infty, \quad \frac{x}{n} \to 0.$$

**Theorem 4.2.** Let  $\mathbf{E}\xi = 0$ ,  $\mathbf{E}\xi^2 =: \sigma^2$ , and condition  $[\mathbf{C}_0]$  holds. Then the sequence of processes  $s_n(t)$  satisfies LDP with speed  $\frac{x^2}{n}$  in space  $(\mathbb{C}, \rho_{\kappa})$  with  $\kappa = 0$ , and rate function  $I_0$ 

$$I_0(f) := \begin{cases} \frac{1}{2\sigma^2} \int_0^\infty (f'(t))^2 dt, & \text{if } f(0) = 0, & f \text{ is absolutely continuous} \\ \infty & \text{otherwise}, \end{cases}$$

i.e. for any measurable set  $B \subset \mathbb{C}$ 

$$\overline{\lim_{n \to \infty}} \, \frac{n}{x^2} \ln \mathbf{P}(s_n \in B) \le -I_0([B]), \quad \underline{\lim_{n \to \infty}} \, \frac{n}{x^2} \ln \mathbf{P}(s_n \in B) \ge -I_0((B)).$$

The proof is similar to that of Theorem 4.1 with replacing n by  $\frac{x^2}{n}$ .

Condition I is verified with help of [17] (Theorem 1) or [6] (Theorem 2.2). Condition II is obvious. Only Condition III requires clarification, which is done by using the following form of Kolmogorov's inequality ([1], p. 295, lemma 11.2.1): for any  $x \ge 0$ ,  $y \ge 0$ ,  $n \ge 1$ 

$$\mathbf{P}(\max_{1 \le m \le n} |S_m| \ge x + y) \le \frac{\mathbf{P}(|S_n| \ge x)}{\min_{1 \le m \le n} \mathbf{P}(|S_m| \le y)}.$$
(4.3)

It is clear that

$$\mathbf{P}\left(\sup_{t\geq T}\frac{|s_{n}(t)|}{1+t}\geq\varepsilon\right) \leq \sum_{K\geq T}\mathbf{P}\left(\sup_{K\leq t\leq K+1}|s_{n}(t)|\geq K\varepsilon\right) \\
\leq \sum_{K\geq T}\mathbf{P}\left(\sup_{K\leq t\leq K+1}|s_{n}(t)-s_{n}(K)|\geq K\varepsilon/2\right) + \sum_{K\geq T}\mathbf{P}\left(|s_{n}(K)|\geq K\varepsilon/2\right) \\
\leq \sum_{K\geq T}P_{1}(K,n) + \sum_{K\geq T}P_{2}(K,n),$$
(4.4)

where

$$P_1(K,n) := \mathbf{P}(\sup_{0 \le t \le 1} |s_n(t)| \ge K\varepsilon/2), \text{ and } P_2(K,n) := \mathbf{P}(|s_n(K)| \ge K\varepsilon/2).$$

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We bound  $P_2(K, n)$  using exponential Chebyshev's (Chernoff) inequality:

$$P_2(K,n) = \mathbf{P}\left(\frac{|S_{nK}|}{nK} \ge \frac{x\varepsilon}{2n}\right) \le e^{-nK\Lambda(\frac{x\varepsilon}{2n})} + e^{-nK\Lambda(-\frac{x\varepsilon}{2n})}$$

It follows by (4.1) that

$$\sum_{K \ge T} P_2(K, n) \le 2 \frac{e^{-\frac{x^2}{n}T\delta_1}}{1 - e^{-\frac{x^2}{n}\delta_1}},$$
(4.5)

as for *n* large enough  $\frac{x\varepsilon}{2n} \leq 1$ , and  $nK\Lambda\left(\pm \frac{x\varepsilon}{2n}\right) \geq \frac{x^2}{n}K\delta_1$ , with  $\delta_1 := \frac{\delta\varepsilon^2}{4}$ . By (4.3)

$$P_1(K,n) = \mathbf{P}\left(\max_{1 \le m \le n} \frac{|S_m|}{xK} \ge \varepsilon/4 + \varepsilon/4\right) \le \frac{\mathbf{P}\left(\frac{|S_n|}{xK} \ge \varepsilon/4\right)}{\min_{1 \le m \le n} \mathbf{P}\left(\frac{|S_m|}{xK} < \varepsilon/4\right)}.$$

Since

$$\min_{\leq m \leq n} \mathbf{P}\left(\frac{|S_m|}{xK} < \varepsilon/4\right) \geq \min_{1 \leq m \leq n} \mathbf{P}\left(\frac{|S_m|}{\sqrt{m}} < \frac{xT}{\sqrt{n}}\varepsilon/4\right) \to 1$$

 $1 \le m \le n \quad \bigvee xK$  as  $n \to \infty$ , for n large enough

$$P_1(K,n) \le 2\mathbf{P}\left(\frac{|S_n|}{xK} \ge \varepsilon/4\right) \le 2e^{-n\Lambda(\frac{xK}{n}\varepsilon/4)} + 2e^{-n\Lambda(-\frac{xK}{n}\varepsilon/4)}$$

It now follows by (4.1)

$$\sum_{K \ge T} P_1(K, n) \le 4 \frac{e^{-\frac{x^2}{n}T\delta_1}}{1 - e^{-\frac{x^2}{n}\delta_1}}.$$
(4.6)

The desired inequality in III follows from (4.5), (4.6) and (4.4): for  $T \geq \frac{N}{\delta_1}$ 

$$\overline{\lim_{n \to \infty}} \frac{n}{x^2} \ln \mathbf{P}\left(\sup_{t \ge T} \frac{|s_n(t)|}{1+t} \ge \varepsilon\right) \le -N.$$

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