

# Asymptotic Behavior for Random Walks in Time-Random Environment on $Z^1$

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**Abstract** In this paper, we give a general model of random walks in time-random environment in any countable space. Moreover, when the environment is independently identically distributed, a recurrence-transience criterion and the law of large numbers are derived in the nearest-neighbor case on  $Z^1$ . At last, under regularity conditions, we prove that the RWIRE  $\{X_n\}$  on  $Z^1$  satisfies a central limit theorem, which is similar to the corresponding results in the case of classical random walks.

**Keywords** Random walks in time-random environment; recurrence-transience criteria; strong law of large numbers; central limit theorem.

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## 1. Introduction

The general model of random walks in space-random environment including a recurrence-transience criterion had been given in Ref. [6]. Moreover, when the environment is i.i.d., the strong law of large numbers was given in Refs. [1] and [4]. However, the limit theorem had also been investigated in Refs. [2] and [5] where the environment is stationary and ergodic. In this paper, we study some asymptotic behavior for random walks in time-random environment (RWIRE) in the nearest-neighbor case on  $Z^1$  as the environment is i.i.d..

We begin with a general setup, that will be specialized later to the cases of interest to us. Now we let  $N = \{0, 1, 2, \dots\}$ . For each  $i \in N$ , let  $M_i(\chi)$  denote the collection of probability measures on  $\chi$  with support  $V$ ,  $V \subset \chi$ , where  $\chi$  is countable. Formally, an element of  $M_i(\chi)$ , called a transition law at time  $i$ , is a measurable function  $\omega_i : \chi \rightarrow [0, 1]$  satisfying:

- $\omega_i(x) \geq 0, \forall x \in V$ ;
- $\omega_i(x) = 0, \forall x \notin V$ ;
- $\sum_{x \in V} \omega_i(x) = 1$ .

We equip  $M_i(\chi)$  with the weak topology on probability measures which makes it become a Polish space. Furthermore, it induces a Polish structure on  $\Omega = \prod_{i \in N} M_i(\chi)$ . Let  $\mathcal{F}$  denote

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the Borel  $\sigma$ -algebra on  $\Omega$ . Given a probability measure  $P$  on  $(\Omega, \mathcal{F})$ . A random environment is an element  $\omega$  of  $\Omega$  distributed according to  $P$ . One defines naturally the shift  $T$  on  $\Omega$  by  $(T\omega)_i = \omega_{i+1}$ ,  $i \in \mathbb{N}$ .

For each  $\omega \in \Omega$ , we define the random walks in the environment  $\omega$  as the space-homogeneous Markov chains  $X = \{X_n, n \geq 0\}$  taking value in  $\chi$  with transition probabilities

$$P_\omega(X_{n+1} = y | X_n = x) = \omega_n(y - x).$$

Fix an environment  $\omega \in \Omega$ ,  $X = \{X_n, n \geq 0\}$  is a time-nonhomogeneous Markov chain. We use  $P_\omega^x$  to denote the law induced on  $(\chi^{\mathbb{N}}, \mathcal{B})$ , where  $\mathcal{B}$  is the  $\sigma$ -algebra generated by cylinder functions and  $P_\omega^x(X_0 = x) = 1$ .

In the sequel, we refer to  $P_\omega^x(\cdot)$  as the quenched law of the random walks  $\{X_n, n \geq 0\}$ . Note that for each  $x \in \chi$  and  $G \in \mathcal{B}$ , the map  $\omega \mapsto P_\omega^x(G)$  is  $\mathcal{F}$ -measurable.

Hence  $P^x := P \otimes P_\omega^x$  on  $(\Omega \times \chi^{\mathbb{N}}, \mathcal{F} \times \mathcal{B})$  is the probability measure defined by

$$P^x(F \times G) = \int_{\mathcal{F}} P_\omega^x(G) P(d\omega), \quad F \in \mathcal{F}, G \in \mathcal{B}.$$

**Example** Let  $\chi = \mathbb{Z}^1$  and  $V = \{-1, 0, 1\}$ . Then according to above definition the RWIRE is called the nearest-neighbor RWIRE on  $\mathbb{Z}^1$ .

## 2. Recurrence-transience criteria

In this section, suppose  $\{X_n, n \geq 0\}$  is the nearest-neighbor RWIRE on  $\mathbb{Z}^1$ , starting at 0. We let  $\omega_n^+ := \omega_n(1)$ ,  $\omega_n^- := \omega_n(-1)$  and  $\omega_n^0 := \omega_n(0)$ .

**Assumption (I)** (a)  $P$  is i.i.d.; (b)  $P\{(\omega_0^+ + \omega_0^-) > 0\} = 1$ .

**Theorem 2.1** Assume Assumption (I). Then

- (1)  $E_P \omega_0^- < E_P \omega_0^+ \implies \lim_{n \rightarrow \infty} X_n = +\infty$ ,
- (2)  $E_P \omega_0^- > E_P \omega_0^+ \implies \lim_{n \rightarrow \infty} X_n = -\infty$ ,
- (3)  $E_P \omega_0^- = E_P \omega_0^+ \implies -\infty = \lim_{n \rightarrow \infty} \inf X_n < \lim_{n \rightarrow \infty} \sup X_n = +\infty$

hold  $P^0$ -a.s., where  $E_P$  is the expectation operator w.r.t.  $P$ .

**Proof** (1) Fix an environment  $\omega$  with  $\frac{E_P \omega_0^-}{E_P \omega_0^+} < \infty$ . For each  $x \in \mathbb{Z}^1$ ,  $l \leq x \leq s$ , define

$$H_{l,s,\omega}^x = P_\omega^x(\{X_n\} \text{ hits } l \text{ before hitting } s).$$

From the assumption  $\frac{E_P \omega_0^-}{E_P \omega_0^+} < \infty$ , for each  $x$ , it follows that  $H_{l,s,\omega}^x$  is well defined as  $P_\omega^x(\{X_n\} \text{ never hits } [l, s]^c) = 0$ . The Markov property of  $P_\omega^x$  implies

$$\begin{cases} H_{l,s,\omega}^x = \omega_0^+ H_{l,s,T\omega}^{x+1} + \omega_0^0 H_{l,s,T\omega}^x + \omega_0^- H_{l,s,T\omega}^{x-1}, & x \in (l, s); \\ H_{l,s,T^k\omega}^l = 1, & k \geq 1; \\ H_{l,s,T^k\omega}^s = 0, & k \geq 1. \end{cases} \quad (2.1)$$

Since  $P$  is i.i.d., and  $P_\omega^x$  is space-homogeneous for all  $\omega \in \Omega$ , it follows by taking expectation on

(2.1) w.r.t.  $P$  that

$$\begin{cases} (E_P\omega_0^+ + E_P\omega_0^-)E_P H_{l,s,\omega}^x = E_P\omega_0^+ E_P H_{l,s,\omega}^{x+1} + E_P\omega_0^- E_P H_{l,s,\omega}^{x-1}, & x \in (l, s), \\ E_P H_{0,s-l,\omega}^0 = 1, \\ E_P H_{l-s,0,\omega}^0 = 0. \end{cases} \quad (2.2)$$

Solving (2.2), we obtain:

(i) If  $E_P\omega_0 \times E_P\omega_0^- > 0$ , then

$$E_P H_{l,s,\omega}^x = \frac{\sum_{j=x+1}^{s-1} \left(\frac{E_P\omega_0^-}{E_P\omega_0^+}\right)^j}{\sum_{j=l+1}^{s-1} \left(\frac{E_P\omega_0^-}{E_P\omega_0^+}\right)^j}, \quad E_P H_{l,s,\omega}^0 = \frac{\sum_{j=1}^{s-1} \left(\frac{E_P\omega_0^-}{E_P\omega_0^+}\right)^j}{\sum_{j=l+1}^{s-1} \left(\frac{E_P\omega_0^-}{E_P\omega_0^+}\right)^j}.$$

There are three cases:

(a) When  $E_P\omega_0^- < E_P\omega_0^+$ , we have  $\lim_{l \rightarrow -\infty} \lim_{s \rightarrow +\infty} E_P H_{l,s,\omega}^0 = 0$ , but  $0 \leq H_{0,s,\omega}^1 \leq 1$ , so  $P(\lim_{l \rightarrow -\infty} \lim_{s \rightarrow +\infty} H_{l,s,\omega}^0 = 0) = 1$ . We also have  $\lim_{s \rightarrow +\infty} E_P H_{-1,s,\omega}^0 < 1$ , hence  $P(\lim_{s \rightarrow +\infty} H_{-1,s,\omega}^0 < 1) > 0$ , which implies  $\lim_{n \rightarrow \infty} X_n = +\infty$  under  $P^0$ -a.s..

(b) When  $E_P\omega_0^- > E_P\omega_0^+$ , similarly, we may get  $\lim_{n \rightarrow \infty} X_n = -\infty$  under  $P^0$ -a.s..

(c) If  $E_P\omega_0^- = E_P\omega_0^+$ , for any fixed  $l$ ,  $\lim_{s \rightarrow +\infty} E_P H_{l,s,\omega}^0 = 1$ , hence  $P(\lim_{s \rightarrow +\infty} H_{l,s,\omega}^0 = 1) = 1$ . Moreover, for any fixed  $s$ ,  $\lim_{l \rightarrow -\infty} E_P H_{l,s,\omega}^0 = 0$ , hence  $P(\lim_{l \rightarrow -\infty} H_{l,s,\omega}^0 = 0) = 1$ , so  $E_P\omega_0^- = E_P\omega_0^+ \implies -\infty = \lim_{n \rightarrow \infty} \inf X_n < \lim_{n \rightarrow \infty} \sup X_n = +\infty$  under  $P^0$ -a.s..

(ii) If  $E_P\omega_0^- \times E_P\omega_0^+ = 0$ , there are two cases by assumption (I)(b):

(a) When  $E_P\omega_0^- = 0, E_P\omega_0^+ > 0$ , then by (2.2) and  $P_\omega$  is space-homogeneous for all  $\omega$

$$E_P(H_{l,s,\omega}^0) = E_P(H_{l-1,s-1,\omega}^0) = \cdots = E_P(H_{l-s,0,\omega}^0) = 0.$$

(b) When  $E_P\omega_0^- > 0, E_P\omega_0^+ = 0$ , similarly, we have

$$E_P(H_{l,s,\omega}^0) = E_P(H_{l+1,s+1,\omega}^0) = \cdots = E_P(H_{0,s-l,\omega}^0) = 1.$$

So the case (a) (or (b)) of (ii) can be included in the case of (a) (or (b)) of (i). We complete the proof by (i) and (ii).  $\square$

### 3. Strong law of large numbers

We introduce hitting times which will serve us later. Let  $T_0 = 0$  and

$$T_n = \inf\{k : X_k = n\}, n \geq 1; \inf \varphi = +\infty.$$

Set  $\tau_0 = 0$  and  $\tau_n = T_n - T_{n-1}, n \geq 1$ . Similarly, set  $T_{-n} = \inf\{k : X_k = -n\}, n \geq 1$  and  $\tau_{-n} = T_{-n} - T_{-n+1}, n \geq 1$ , with the convention that  $\tau_{\pm n} = +\infty$  if  $T_{\pm n} = \infty$ .

**Theorem 3.1** Assume Assumption (I). Then under  $P^0$ -a.s.

- (1)  $E_P\omega_0^+ > E_P\omega_0^- \implies \lim_{n \rightarrow \infty} \frac{T_n}{n} = (E_P\omega_0^+ - E_P\omega_0^-)^{-1}, \lim_{n \rightarrow \infty} \frac{X_n}{n} = (E_P\omega_0^+ - E_P\omega_0^-);$
- (2)  $E_P\omega_0^+ < E_P\omega_0^- \implies \lim_{n \rightarrow \infty} \frac{T_{-n}}{n} = (E_P\omega_0^- - E_P\omega_0^+)^{-1}, \lim_{n \rightarrow \infty} \frac{X_n}{n} = (E_P\omega_0^+ - E_P\omega_0^-);$
- (3)  $E_P\omega_0^+ = E_P\omega_0^- \implies \lim_{n \rightarrow \infty} \frac{T_n}{n} = \infty = \lim_{n \rightarrow \infty} \frac{T_{-n}}{n}, \lim_{n \rightarrow \infty} \frac{X_n}{n} = 0.$

For the proof of Theorem 3.1, we need the following two lemmas.

**Lemma 3.1** Assume Assumption (I). Then  $\{\tau_n, n \geq 1\}$  is i.i.d. and  $\{\tau_{-n}, n \geq 1\}$  is i.i.d. under the law  $P^0$ -a.s..

**Proof** When  $E_P\omega_0^+ \geq E_P\omega_0^-$ , we have  $P^0(\lim_{n \rightarrow \infty} \sup X_n = +\infty) = 1$ , by the definition of  $\tau_n$ ,  $P^0(\tau_n < \infty) = 1$  for all  $n \geq 1$ . To prove  $\{\tau_n\}$  is i.i.d., it suffices to show for any positive integer  $k$  and  $m$ ,

$$P^0(\tau_2 = k | \tau_1 = m) = P^0(\tau_2 = k) = P^0(\tau_1 = k).$$

For any fixed  $\omega \in \Omega$ , by the Markov property and the space-homogeneity of  $P_\omega^0$ , we have

$$P_\omega^0(\tau_2 = k) = \sum_{m=1}^{\infty} P_\omega^0(\tau_2 = k | \tau_1 = m) P_\omega^0(\tau_1 = m) = \sum_{m=1}^{\infty} P_{T^m\omega}^0(\tau_1 = k) P_\omega^0(\tau_1 = m). \quad (3.1)$$

Since  $P$  is i.i.d., it follows by taking expectation on (3.1) w.r.t.  $P$  that

$$P^0(\tau_2 = k) = P \otimes P_\omega^0(\tau_1 = k) \sum_{m=1}^{\infty} P \otimes P_\omega^0(\tau_1 = m) = P^0(\tau_1 = k). \quad (3.2)$$

On the other hand

$$\begin{aligned} P_\omega^0(\tau_2 = k | \tau_1 = m) &= P_\omega^0(X_{m+k} = 2, X_{m+s} \neq 2, s = 1, 2, \dots, k-1 | \\ X_m = 1, X_n \neq 1, n = 1, 2, \dots, m-1) &= P_{T^m\omega}^0(\tau_1 = k). \end{aligned} \quad (3.3)$$

Hence by taking expectation on (3.3) w.r.t.  $P$ , we also have

$$P^0(\tau_2 = k | \tau_1 = m) = P^0(\tau_1 = k). \quad (3.4)$$

So  $\{\tau_n, n \geq 1\}$  is i.i.d. under  $P^0$ -a.s. by (3.2) and (3.4) when  $E_P\omega_0^+ \leq E_P\omega_0^-$ . Similarly, we may show  $\{\tau_{-n}, n \geq 1\}$  is i.i.d. under  $P^0$ -a.s..

**Lemma 3.2** Assume Assumption (I). Then

$$\begin{aligned} (1) \quad E_{P^0}(\tau_1) &= \begin{cases} (E_P\omega_0^+ - E_P\omega_0^-)^{-1}, & E_P\omega_0^+ > E_P\omega_0^-, \\ +\infty, & E_P\omega_0^+ = E_P\omega_0^-; \end{cases} \\ (2) \quad E_{P^0}(\tau_{-1}) &= \begin{cases} (E_P\omega_0^- - E_P\omega_0^+)^{-1}, & E_P\omega_0^- > E_P\omega_0^+, \\ +\infty, & E_P\omega_0^- = E_P\omega_0^+. \end{cases} \end{aligned}$$

**Proof** We prove only (1) since the proof of (2) is similar. Decompose, with  $X_0 = 0$

$$\tau_1 = \chi_{\{X_1=1\}} + \chi_{\{X_1=0\}}(1 + \tau_1') + \chi_{\{X_1=-1\}}(1 + \tau_0'' + \tau_1''). \quad (3.5)$$

Here  $\tau_1'$  is the first hitting time of 1 after time 1 (possible infinite),  $1 + \tau_0''$  is the first hitting time of 0 after time 1, and  $1 + \tau_0'' + \tau_1''$  is the first hitting time of 1 after time  $1 + \tau_0''$ .

Consider first the case  $E_{P^0}\tau_1 < \infty$ . Then  $E_\omega^0\tau_1 < \infty$  under  $P$ -a.s.. Taking expectation on (3.5), one gets

$$\begin{aligned} E_\omega^0\tau_1 &= P_\omega^0(X_1 = 1) + P_\omega^0(X_1 = 0)(1 + E_{T\omega}^0\tau_1) + \\ &P_\omega^0(X_1 = -1)(1 + E_{T\omega}^0\tau_1 + \sum_{m=1}^{\infty} P_{T^m\omega}^0(\tau_1 = m)E_{T^{m+1}\omega}^0\tau_1) \end{aligned}$$

$$= 1 + (1 - \omega_0^+) E_{T\omega}^0 \tau_1 + \omega_0^- \sum_{m=1}^{\infty} P_{T\omega}^0(\tau_1 = m) E_{T^{m+1}\omega}^0 \tau_1. \quad (3.6)$$

Taking expectation on (3.6) w.r.t.  $P$ , we get that

$$E_{P^0} \tau_1 = 1 + (1 - E_P \omega_0^+) E_{P^0} \tau_1 + E_P \omega_0^- E_{P^0} \tau_1. \quad (3.7)$$

Hence when  $E_P \omega_0^+ > E_P \omega_0^-$ , we have  $E_{P^0} \tau_1 = (E_P \omega_0^+ - E_P \omega_0^-)^{-1} < \infty$ .

Note next that if  $E_P \omega_0^+ \geq E_P \omega_0^-$ , we have by Theorem 2.1 that  $E_P(\tau_1 \chi_{\tau_1 < \infty}) = E_P \tau_1$ . Hence  $E_{P^0} \tau_1 = \infty$  implies  $E_P \omega_0^+ = E_P \omega_0^-$ ; on the other hand, if  $E_P \omega_0^+ = E_P \omega_0^-$ , by (3.7) we also have  $E_{P^0} \tau_1 = \infty$ . We finish the proof of Lemma 3.2.

**Proof of Theorem 3.1** An application of Lemmas 3.1 and 3.2 yields that in case (1)

$$\frac{T_n}{n} = \frac{\sum_{i=1}^n \tau_i}{n} \longrightarrow E_{P^0} \tau_1 < \infty. \quad (3.8)$$

Similarly, we use  $-n$  instead of  $n$ , we also get that in case (2)

$$\frac{T_{-n}}{n} = \frac{\sum_{i=1}^n \tau_{-i}}{n} \longrightarrow E_{P^0} \tau_{-1} < \infty. \quad (3.9)$$

However when  $E_P \omega_0^+ = E_P \omega_0^-$ ,  $P^0(-\infty = \lim_{n \rightarrow \infty} \inf X_n < \lim_{n \rightarrow \infty} \sup X_n = +\infty) = 1$ , for  $n$  large enough,  $T_n = +\infty = T_{-n}$ , so in case (3),

$$P^0\left(\lim_{n \rightarrow \infty} \frac{T_n}{n} = +\infty = \lim_{n \rightarrow \infty} \frac{T_{-n}}{n}\right) = 1.$$

Now we prove the second limit of the case (1). Let  $K_n$  be the unique (random) integers such that  $T_{K_n} \leq n < T_{K_n+1}$ . Note that  $X_n \leq K_n + 1$ , while  $X_n \geq K_n - (n - T_{K_n})$ . Hence

$$\frac{K_n}{n} - \left(1 - \frac{T_{K_n}}{n}\right) \leq \frac{X_n}{n} < \frac{K_n + 1}{n}. \quad (3.10)$$

Since  $P^0(\lim_{n \rightarrow \infty} X_n = +\infty) = 1$ , from the definition of  $K_n$ ,  $P^0(\lim_{n \rightarrow \infty} \frac{T_{K_n}}{n} = 1) = 1$ . But the definition  $K_n$  also implies

$$P^0\left(\lim_{n \rightarrow \infty} \frac{K_n}{n} = \lim_{n \rightarrow \infty} \frac{n}{T_n}\right) = 1.$$

Thus it follows from (3.8) and (3.10) that

$$P^0\left(\lim_{n \rightarrow \infty} \frac{X_n}{n} = \lim_{n \rightarrow \infty} \frac{n}{T_n} = E_P \omega_0^+ - E_P \omega_0^-\right) = 1.$$

Similarly, we may prove the second limit of (2) and (3).  $\square$

#### 4. The central limit Theorem

In this section, we study the limiting distribution of the RWIRW  $\{X_n\}$ , we use following notations:

$$\begin{aligned} \mu &= (E_{P^0} \tau_1)^{-1}, \sigma^2 = \frac{E_{P^0}(\tau_1)^2 - (E_{P^0} \tau_1)^2}{(E_{P^0} \tau_1)^3}, D_{P^0} \tau_1 = E_{P^0}(\tau_1)^2 - (E_{P^0} \tau_1)^2; \\ \mu_{-1} &= (E_{P^0} \tau_{-1})^{-1}, \sigma_{-1}^2 = \frac{E_{P^0}(\tau_{-1})^2 - (E_{P^0} \tau_{-1})^2}{(E_{P^0} \tau_{-1})^3}, D_{P^0} \tau_{-1} = E_{P^0}(\tau_{-1})^2 - (E_{P^0} \tau_{-1})^2. \end{aligned}$$

**Theorem 4.1** Assume Assumption (I), under  $P^0$ -a.s.,

(1) If  $E_P\omega_0^+ > E_P\omega_0^-$  and  $D_{P^0}\tau_1 < \infty$ , then

$$\frac{X_n - n\mu}{\sigma\sqrt{n}} \longrightarrow N(0, 1); \quad \frac{T_n - n\mu^{-1}}{\sqrt{nD_{P^0}\tau_1}} \longrightarrow N(0, 1), \quad n \rightarrow \infty.$$

(2) If  $E_P\omega_0^+ = E_P\omega_0^-$ , then

$$\frac{X_n}{\sqrt{2nE_P\omega_0^+}} \longrightarrow N(0, 1), \quad n \rightarrow \infty.$$

(3) If  $E_P\omega_0^+ < E_P\omega_0^-$  and  $D_{P^0}\tau_{-1} < \infty$ , then

$$\frac{X_n - n\mu_{-1}}{\sigma_{-1}\sqrt{n}} \longrightarrow N(0, 1); \quad \frac{T_n - n\mu_{-1}^{-1}}{\sqrt{nD_{P^0}\tau_{-1}}} \longrightarrow N(0, 1), \quad n \rightarrow \infty.$$

**Proof** (1) For any positive integers  $n$ ,  $L$  and  $M$ , by the definition of  $T_n$ , we have

$$\begin{aligned} \{T_L \geq n\} &\subset \{X_n \leq L\} \subset (\{T_{L+M} \geq n\} \cup \{(\inf_{s \geq T_{L+M}} X_s \leq L) \cap (T_{L+M} < n)\}) \\ &\subset (\{T_{L+M} \geq n\} \cup \{\inf_{s \geq T_{L+M}} X_s - (L+M) \leq -M\}). \end{aligned} \quad (4.1)$$

Since  $P$  is i.i.d., it follows by the Markov property and the space-homogeneity of  $P_\omega^0$  that

$$P^0(\inf_{s \geq T_{L+M}} X_s - (L+M) \leq -M) = P^0(\inf_{s \geq 0} X_s \leq -M). \quad (4.2)$$

By Theorem 2.1 we have

$$\lim_{M \rightarrow \infty} P^0(\inf_{s \geq 0} X_s \leq -M) = 0. \quad (4.3)$$

Hence for any positive integers  $l$  and for any  $\delta > 0$ , there exists  $M$  large enough such that

$$P^0(T_L \geq n) \leq P^0(X_n \leq L) \leq P^0(T_{L+M} \geq n) + \delta. \quad (4.4)$$

Now for any given real  $x$ , take

$$L = L(n, x) = n\mu + x\sigma\sqrt{n} + o(\sqrt{n}), \quad (4.5)$$

we have  $L(n, x) \longrightarrow +\infty$  as  $n \rightarrow \infty$  and

$$P^0(X_n \leq L(n, x)) \approx P^0\left(\frac{X_n - n\mu}{\sigma\sqrt{n}} \leq x\right). \quad (4.6)$$

On the other hand

$$P^0(T_{L(n, x)} \geq n) = P^0\left(\frac{T_{L(n, x)} - L(n, x)\mu^{-1}}{\sqrt{L(n, x)D_{P^0}\tau_1}} \geq \frac{n - L(n, x)\mu^{-1}}{\sqrt{L(n, x)D_{P^0}\tau_1}}\right).$$

By the definition of  $L(n, x)$ , we have

$$\lim_{n \rightarrow \infty} \frac{n - L(n, x)\mu^{-1}}{\sqrt{L(n, x)D_{P^0}\tau_1}} = -x.$$

By the Central Limit Theorem of i.i.d. random variable sequence and the given condition we obtain

$$\lim_{n \rightarrow \infty} P^0(T_{L(n, x)} \geq n) = \lim_{n \rightarrow \infty} P^0(T_{L(n, x)+M} \geq n) = 1 - \Phi(-x) = \Phi(x),$$

where  $\Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x \exp(-\frac{t^2}{2}) dt$ . So by (4.4) and (4.6), the case (1) is proved. Similarly, we can prove the case (3).

(2) When  $E_P\omega_0^+ = E_P\omega_0^-$ , suppose  $X_n = \sum_{i=1}^n Y_i, Y_i \in \{-1, 0, 1\}$ , where

$$E_\omega(Y_i = 1) = \omega_i^+, E_\omega(Y_i = -1) = \omega_i^-, E_\omega(Y_i = 0) = \omega_i^0.$$

Since  $P$  is i.i.d., by the given condition we have

$$\begin{aligned} E_{P^0}X_n &= E_P\left[\sum_{i=1}^n (\omega_i^+ - \omega_i^-)\right] = \sum_{i=1}^n (E_P\omega_i^+ - E_P\omega_i^-) = 0, \\ E_{P^0}X_n^2 &= E_P(E_\omega^0(\sum_{i=1}^n \sum_{j=1}^n Y_i Y_j)) = E_P(E_\omega^0(\sum_{i=1}^n Y_i^2 + 2 \sum_{1 \leq i < j \leq n} Y_i Y_j)) \\ &= E_P(\sum_{i=1}^n (\omega_i^+ + \omega_i^-) + 2 \sum_{1 \leq i < j \leq n} (\omega_i^+ - \omega_i^-)(\omega_j^+ - \omega_j^-)) = 2 \sum_{i=1}^n E_P\omega_i^+ = 2nE_P\omega_0^+. \end{aligned}$$

Suppose that the characteristic function of  $\frac{X_n}{\sqrt{2nE_P\omega_0^+}}$  is  $\varphi_n(t)$ . Then

$$\begin{aligned} \varphi_n(t) &= E_{P^0} \exp(it \frac{X_n}{\sqrt{2nE_P\omega_0^+}}) = E_P(E_\omega^0 \prod_{j=1}^n \exp(it \frac{Y_j}{\sqrt{2nE_P\omega_0^+}})) \\ &= E_P[\prod_{j=1}^n (\exp(\frac{it}{\sqrt{2nE_P\omega_0^+}})\omega_j^+ + \exp(\frac{-it}{\sqrt{2nE_P\omega_0^+}})\omega_j^- + \omega_j^0)] \\ &= \prod_{j=1}^n [\exp(\frac{it}{\sqrt{2nE_P\omega_0^+}})E_P\omega_j^+ + \exp(\frac{-it}{\sqrt{2nE_P\omega_0^+}})E_P\omega_j^- + E_P\omega_j^0] \\ &= [E_P\omega_0^+ (\exp(\frac{it}{\sqrt{2nE_P\omega_0^+}}) + \exp(\frac{-it}{\sqrt{2nE_P\omega_0^+}})) + E_P\omega_0^0]^n \\ &= [E_P\omega_0^+ (1 + \frac{it}{\sqrt{2nE_P\omega_0^+}} - \frac{1}{2} \cdot \frac{t^2}{2nE_P\omega_0^+} + o(\frac{1}{n})) + 1 - \\ &\quad \frac{it}{\sqrt{2nE_P\omega_0^+}} - \frac{1}{2} \cdot \frac{t^2}{2nE_P\omega_0^+} + o(\frac{1}{n})) + E_P\omega_0^0]^n \\ &= [E_P\omega_0^+ (2 - \frac{t^2}{2nE_P\omega_0^+} + o(\frac{1}{n})) + E_P\omega_0^0]^n \\ &= [1 - \frac{t^2}{2n} + o(\frac{1}{n})]^n \longrightarrow \exp(-\frac{t^2}{2}), \quad n \longrightarrow \infty. \end{aligned} \tag{4.7}$$

It follows by the Continuous Theorem and (4.7) that  $\frac{X_n}{\sqrt{2nE_P\omega_0^+}} \rightarrow N(0, 1)$  as  $n \rightarrow \infty$ .

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