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The quenched limiting distributions of a one-dimensional random walk in random scenery

Nadine Guillotin-Plantard* and Yueyun Hu†

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Summary. For a one-dimensional random walk in random scenery (RWRS) on \mathbb{Z} , we determine its quenched weak limits by applying Strassen [14]’s functional law of iterated logarithm. As a consequence, conditioned on the random scenery, the one-dimensional RWRS does not converge in law, in contrast with the multi-dimensional case.

Keywords: Random walk in random scenery; Weak limit theorem; Law of the iterated logarithm; Brownian motion in Brownian Scenery; Strong approximation.

AMS Subject Classification: 60F05, 60G52.

1. INTRODUCTION

Random walks in random sceneries were introduced independently by Kesten and Spitzer [9] and by Borodin [3, 4]. Let $S = (S_n)_{n \geq 0}$ be a random walk in \mathbb{Z}^d starting at 0, i.e., $S_0 = 0$ and $(S_n - S_{n-1})_{n \geq 1}$ is a sequence of i.i.d. \mathbb{Z}^d -valued random variables. Let $\xi = (\xi_x)_{x \in \mathbb{Z}^d}$ be a field of i.i.d. real random variables independent of S . The field ξ is called the random scenery. The random walk in random scenery (RWRS) $K := (K_n)_{n \geq 0}$ is defined by setting $K_0 := 0$ and, for $n \in \mathbb{N}^*$,

$$K_n := \sum_{i=1}^n \xi_{S_i}. \quad (1.1)$$

We will denote by \mathbb{P} the joint law of S and ξ . The law \mathbb{P} is called the *annealed* law, while the conditional law $\mathbb{P}(\cdot | \xi)$ is called the *quenched* law.

Limit theorems for RWRS have a long history, we refer to [6] or [7] for a complete review. Distributional limit theorems for *quenched* sceneries (i.e. under the quenched law) are however quite recent. The first result in this direction that we are aware of was obtained by Ben Arous and Černý [1], in the case of a heavy-tailed scenery and planar random walk. In [6], quenched central limit theorems (with the usual \sqrt{n} -scaling and Gaussian law in the limit) were proved for a large class of transient random walks. More recently, in [7], the case of the planar random walk was studied, the authors proved a quenched version of the annealed central limit theorem obtained by Bolthausen in [2].

In this note we consider the case of the simple symmetric random walk $(S_n)_{n \geq 0}$ on \mathbb{Z} , the random scenery $(\xi_x)_{x \in \mathbb{Z}}$ is assumed to be centered with finite variance equal to one and there

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exists some $\delta > 0$ such that $\mathbb{E}(|\xi_0|^{2+\delta}) < \infty$. We prove that under these assumptions, there is no quenched distributional limit theorem for K .

Theorem 1. *For \mathbb{P} -a.e. ξ , under the quenched probability $\mathbb{P}(\cdot | \xi)$, the process*

$$\tilde{K}_n := \frac{K_n}{(2n^{3/2} \log \log n)^{1/2}}, \quad n > e^e,$$

does not converge in law. More precisely, for \mathbb{P} -a.e. ξ , under the quenched probability $\mathbb{P}(\cdot | \xi)$, the limit points of the law of \tilde{K}_n , as $n \rightarrow \infty$, under the topology of weak convergence of measures, are equal to the closure of the laws of random variables in \mathcal{K}_B , with

$$\mathcal{K}_B := \left\{ \int_{m_1}^{M_1} f(x) dL_1(x) : f \in \mathcal{C}([m_1, M_1] \rightarrow \mathbb{R}) \text{ such that } f(0) = 0, \int_{m_1}^{M_1} (\dot{f}(x))^2 dx \leq 1 \right\}, \quad (1.2)$$

where $(L_1(x), x \in \mathbb{R})$ denotes the family of local times at time 1 of a one-dimensional Brownian motion B starting from 0 and $m_1 := \inf_{0 \leq s \leq 1} B_s$, $M_1 := \sup_{0 \leq s \leq 1} B_s$.

Let us mention that the random set \mathcal{K}_B directly comes from Strassen [14]'s limiting set. Moreover, \dot{f} is the derivative of the (random) function f and the precise meaning of $\int_{m_1}^{M_1} f(x) dL_1(x)$ can be given by the integration by parts and the occupation times formulas: $\int_{m_1}^{M_1} f(x) dL_1(x) = - \int_{-\infty}^{\infty} L_1(x) \dot{f}(x) dx = - \int_0^1 \dot{f}(B_s) ds$.

Instead of Theorem 1, we shall prove that there is no quenched limit theorem for the continuous analogue of K introduced by Kesten and Spitzer [9] and deduce Theorem 1 by using a strong approximation for the one-dimensional RWRS. Let us define this continuous analogue: Assume that $B := (B(t))_{t \geq 0}$, $W := (W(t))_{t \geq 0}$, $\tilde{W} := (\tilde{W}(t))_{t \geq 0}$ are three real Brownian motions starting from 0, defined on the same probability space and independent of each other. For brevity, we shall write $W(x) := W(x)$ if $x \geq 0$ and $\tilde{W}(-x)$ if $x < 0$ and say that W is a two-sided Brownian motion. We denote by $\mathbb{P}_B, \mathbb{P}_W$ the law of these processes. We will also denote by $(L_t(x))_{t \geq 0, x \in \mathbb{R}}$ a continuous version with compact support of the local time of the process B . We define the continuous version of the RWRS, also called *Brownian motion in Brownian scenery*, as

$$Z_t := \int_0^{+\infty} L_t(x) dW(x) + \int_0^{+\infty} L_t(-x) d\tilde{W}(x) \equiv \int_{-\infty}^{+\infty} L_t(x) dW(x).$$

It was proved by Kesten and Spitzer [9] that the process $(n^{-3/4} K([nt]))_{t \geq 0}$ weakly converges in the space of continuous functions to the continuous process $Z = (Z(t))_{t \geq 0}$. Zhang [15] (see also [5, 10]) gave a stronger version of this result in the special case when the scenery has a finite moment of order $2 + \delta$ for some $\delta > 0$, more precisely, there is a coupling of ξ , S , B and W such that (ξ, W) is independent of (S, B) and for any $\varepsilon > 0$, almost surely,

$$\max_{0 \leq m \leq n} |K(m) - Z(m)| = o(n^{\frac{1}{2} + \frac{1}{2(2+\delta)} + \varepsilon}), \quad n \rightarrow +\infty. \quad (1.3)$$

Theorem 1 will follow from this strong approximation and the following result.

Theorem 2. *\mathbb{P}_W -almost surely, under the quenched probability $\mathbb{P}(\cdot | W)$, the limit points of the law of*

$$\tilde{Z}_t := \frac{Z_t}{(2t^{3/2} \log \log t)^{1/2}}, \quad t \rightarrow \infty,$$

under the topology of weak convergence of measures, are equal to the closure of the laws of random variables in \mathcal{K}_B defined in Theorem 1. Consequently under $\mathbb{P}(\cdot | W)$, as $t \rightarrow \infty$, \tilde{Z}_t does not converge in law.

To prove Theorem 2, we shall apply Strassen [14]’s functional law of iterated logarithm applied to the two-sided Brownian motion W ; we shall also need to estimate the stochastic integral $\int g_{m_1, M_1}(x) dL_1(x)$ for a bounded Borel function g , this will be done by using the theory of enlargement of filtrations, see Section 2 for the details.

2. PROOFS

At first, we reformule Strassen’s theorem as follows: For a two-sided one-dimensional Brownian motion $(W(t), t \in \mathbb{R})$ starting from 0, define for any $\lambda > e^e$,

$$W_\lambda(t) := \frac{W(\lambda t)}{(2\lambda \log \log \lambda)^{1/2}}, \quad t \in \mathbb{R}.$$

Theorem 3 (Strassen). *Almost surely, for any $s < 0 < r$, $(W_\lambda(t), s \leq t \leq r)$ is relatively compact in the uniform topology and the set of its limit points is $\mathcal{K}_{s,r}$, with*

$$\mathcal{K}_{s,r} := \left\{ f \in \mathcal{C}([s, r] \rightarrow \mathbb{R}) : f(0) = 0, \int_s^r (\dot{f}(x))^2 dx \leq 1 \right\}.$$

Proof: For any fixed $s < 0 < r$, by applying Strassen’s theorem ([14]) to the two-dimensional rescaled Brownian motion: $(\frac{W(\lambda r u)}{\sqrt{2\lambda r \log \log \lambda}}, \frac{W(\lambda |s| u)}{\sqrt{2\lambda |s| \log \log \lambda}})_{0 \leq u \leq 1}$, we get that $(W_\lambda(t), s \leq t \leq r)$ is relatively compact in the uniform topology with $\mathcal{K}_{s,r}$ as the set of limit points. By inverting a.s. and s, r , we see that there exists an event Ω_0 with full probability such that for any $\omega \in \Omega_0$, for all $s < 0 < r$ integers, $(W_\lambda(t)(\omega), s \leq t \leq r)$ is relatively compact in the uniform topology and the set of its limit points is $\mathcal{K}_{s,r}$.

Now, let $\omega \in \Omega_0$. We are going to check that for any $s < 0 < r$, $(W_\lambda(t)(\omega), s \leq t \leq r)$ is relatively compact in the uniform topology and the set of its limit points is $\mathcal{K}_{s,r}$.

To this end, for any $s < 0 < r$, let $k < 0 < l$ be integers such that $k \leq s < k+1 \leq 0 \leq l-1 < r \leq l$. Let $d(\cdot, \cdot)$ be the distance in the uniform topology: $d(f, g) := \sup_{a \leq t \leq b} |f(t) - g(t)|$ for $f, g \in \mathcal{C}([a, b] \rightarrow \mathbb{R})$, which is consistent on $a < 0 < b$. For any $\varepsilon > 0$, for all large $\lambda \geq \lambda_0(\varepsilon, \omega)$, $d(W_\lambda(\cdot)(\omega), \mathcal{K}_{k,l}) < \varepsilon$, hence there exists some $f \equiv f_{\lambda, \omega, k, l} \in \mathcal{K}_{k,l}$ such that $d(W_\lambda(\cdot)(\omega), f) < \varepsilon$. Notice that if we denote by $g := f1_{[s, r]}(\cdot)$ the restriction of f on $[s, r]$, then $g \in \mathcal{K}_{s,r}$ and $d(W_\lambda(\cdot)(\omega), g) < \varepsilon$, this proves that for any $\omega \in \Omega_0$, all possible limits of $(W_\lambda(t)(\omega), s \leq t \leq r)$ as $\lambda \rightarrow \infty$ are in $\mathcal{K}_{s,r}$.

It remains to check that for any $f \in \mathcal{K}_{s,r}$, infinitely often $\lambda \rightarrow \infty$, $(W_\lambda(t)(\omega), s \leq t \leq r)$ are in the ball centered at f of radius ε . It is enough to check this for all $f \in \mathcal{K}_{s,r}$ such that $\int_s^r |\dot{f}(t)|^2 dt < 1$. We may find some function $h \in \mathcal{K}_{k,l}$ such that f is the restriction of h on $[s, r]$. Since infinitely often as $\lambda \rightarrow \infty$, $(W_\lambda(t)(\omega), k \leq t \leq l)$ are in the ball centered at h of radius ε , the desired conclusion follows. \square

Next, we recall some properties on Brownian local times: The process $x \rightarrow L_1(x)$ is a (continuous) semimartingale (by Perkins [12]), moreover, the quadratic variation of $x \rightarrow L_1(x)$ equals $4 \int_{-\infty}^x L_1(z) dz$. By Revuz and Yor ([13], Exercice VI (1.28)), for any f locally bounded Borel function,

$$\frac{1}{2} \int f(x) dL_1(x) = - \int_0^{B_1} f(u) du + \int_0^1 f(B_u) dB_u. \quad (2.4)$$

Recall that $m_1 = \inf_{0 \leq t \leq 1} B_t$ and $M_1 = \sup_{0 \leq t \leq 1} B_t$. When f is random but measurable with respect to $\sigma\{m_1, M_1\}$, we define $\int f(x) dL_1(x)$ through (2.4): the integral $\int_0^1 f(B_u) dB_u$ is well defined since by the theory of (initial) enlargement of filtration (see e.g. Jeulin and Yor [8], Mansuy and Yor [11]), B is still a continuous semimartingale with respect to the filtration

$(\sigma\{B_s, s \leq t, m_1, M_1\})_{0 \leq t \leq 1}$ (see the forthcoming (2.11) for the semimartingale decomposition). Denote by $L^1(B) := L^1(\mathbb{P}_B, \sigma\{B_s, 0 \leq s \leq 1\})$ and $\|\cdot\|_1$ the L^1 -norm.

Lemma 4. *There exists some constant $c_1 > 0$ such that for any bounded Borel function $g : (a, b, x) \in \mathbb{R}^3 \rightarrow g_{a,b}(x) \in \mathbb{R}$, we have*

$$\left\| \int g_{m_1, M_1}(x) dL_1(x) \right\|_1 \leq c_1 \sup_{a,b,x} |g_{a,b}(x)|.$$

Proof: Let $\|g\|_\infty := \sup_{a,b,x} |g_{a,b}(x)|$. Using (2.4) and observing that $\|\int_0^{B_1} g_{m_1, M_1}(u) du\|_1 \leq \|g\|_\infty \mathbb{E}[|B_1|]$, it is enough to prove that for some positive constant c_2 ,

$$\left\| \int_0^1 g_{m_1, M_1}(B_u) dB_u \right\|_1 \leq c_2 \|g\|_\infty. \quad (2.5)$$

The difficulty here is the dependence of g on (m_1, M_1) , in particular, there is no L^2 -isometry for $\int_0^1 g_{m_1, M_1}(B_u) dB_u$. We need here an (initial) enlargement of filtration (see e.g. Jeulin and Yor [8], Mansuy and Yor [11]): Let (\mathcal{F}_t) be the natural filtration of B and $\tilde{\mathcal{F}}_t := \sigma\{M_1, m_1\} \vee \mathcal{F}_t$ for $0 \leq t \leq 1$. We compute at first the law of (m_1, M_1) conditioned on \mathcal{F}_t : for any measurable bounded function $h : \mathbb{R}^2 \rightarrow \mathbb{R}$ and $0 < t < 1$,

$$\mathbb{E}\left(h(m_1, M_1) | \mathcal{F}_t\right) = \mathbb{E}\left(h(m_t \wedge (B_t + \widehat{m}_{1-t}), M_t \vee (B_t + \widehat{M}_{1-t})) | \mathcal{F}_t\right),$$

where $(\widehat{m}_{1-t}, \widehat{M}_{1-t})$ are independent of \mathcal{F}_t and have the same law as (m_{1-t}, M_{1-t}) , and $x \wedge y := \min(x, y)$, $x \vee y := \max(x, y)$ for any $x, y \in \mathbb{R}$. For any fixed $r > 0$, the law of (m_r, B_r, M_r) is given as follows (cf. Revuz and Yor [13], Exercice III (3.15)): For any $a < 0 < b$,

$$\mathbb{P}\left(a \leq m_r < M_r \leq b, B_r \in dx\right)/dx = \sum_{k=-\infty}^{\infty} \left(p_r(x+2k(b-a)) - p_r(x-2b+2k(b-a))\right) =: \Upsilon_{r,a,b}(x), \quad (2.6)$$

with $p_r(x) := (2\pi r)^{-1/2} e^{-x^2/(2r)}$. Integrating on $x \in [a, b]$ gives that

$$\phi_r(a, b) := \mathbb{P}\left(a \leq m_r < M_r \leq b\right) = \sum_{k=-\infty}^{\infty} (-1)^k \int_{a+k(b-a)}^{b+k(b-a)} p_r(x) dx. \quad (2.7)$$

Therefore for any fixed $r > 0$,

$$\mathbb{P}\left(m_r \in da, M_r \in db\right) = \psi_r(a, b) da db, \quad a < 0 < b,$$

with

$$\begin{aligned} \psi_r(a, b) &:= \sum_{k=-\infty}^{\infty} (-1)^k \left(k(k+1) p'_r(b+k(b-a)) - k(k-1) p'_r(a+k(b-a)) \right) \\ &= 2 \sum_{k=-\infty}^{\infty} (-1)^k k(k+1) p'_r(b+k(b-a)), \end{aligned} \quad (2.8)$$

where $p'_r(x) = -\frac{x}{r} p_r(x)$ denotes the derivative of $p_r(x)$ with respect to x . Plainly, $\psi_r(a, b)$ is continuous on (a, b, r) for $a < 0 < b$ and $r > 0$. We can show that $\psi_r(a, b) > 0$ for any $a < 0 < b$ and $r > 0$, see Remark 5.

It follows that for any $0 < t < 1$,

$$\begin{aligned}\mathbb{E}\left(h(m_1, M_1)|\mathcal{F}_t\right) &= \int_{a < 0 < b} da db \psi_{1-t}(a, b) h(m_t \wedge (B_t + a), M_t \vee (B_t + b)) \\ &= I_1 + I_2 + I_3 + I_4,\end{aligned}$$

with

$$\begin{aligned}I_1 &:= \int_{-\infty}^{m_t} dx \int_{M_t}^{\infty} dy \psi_{1-t}(x - B_t, y - B_t) h(x, y), \\ I_2 &:= \int_{-\infty}^{m_t} dx \int_0^{M_t - B_t} db \psi_{1-t}(x - B_t, b) h(x, M_t), \\ I_3 &:= \int_{m_t - B_t}^0 da \int_0^{M_t - B_t} db \psi_{1-t}(a, b) h(m_t, M_t) = \phi_{1-t}(m_t - B_t, M_t - B_t) h(m_t, M_t), \\ I_4 &:= \int_{m_t - B_t}^0 da \int_{M_t}^{\infty} dy \psi_{1-t}(a, y - B_t) h(m_t, y).\end{aligned}$$

Let us adopt the following notation: for any $f \in C^1((-\infty, 0) \times (0, \infty) \rightarrow \mathbb{R})$,

$$\nabla f(a, b) := \frac{\partial f}{\partial a} + \frac{\partial f}{\partial b}.$$

Define for $a < 0 < b$,

$$\psi_r^{(1)}(a, b) := \int_a^0 dx \psi_r(x, b), \quad \psi_r^{(2)}(a, b) := \int_0^b \psi_r(a, y) dy. \quad (2.9)$$

Then $I_2 = \int_{-\infty}^{m_t} dx \psi_{1-t}^{(2)}(x - B_t, M_t - B_t) h(x, M_t)$ and by Ito's formula,

$$dI_2 = - \left[\int_{-\infty}^{m_t} dx \nabla \psi_{1-t}^{(2)}(x - B_t, M_t - B_t) h(x, M_t) \right] dB_t + \text{f.v.},$$

where here and in the sequel, f.v. denotes some finite variation process. Similarly,

$$\begin{aligned}dI_1 &= - \left[\int_{-\infty}^{m_t} dx \int_{M_t}^{\infty} dy \nabla \psi_{1-t}(x - B_t, y - B_t) h(x, y) \right] dB_t + \text{f.v.}, \\ dI_3 &= - \left[\nabla \phi_{1-t}(m_t - B_t, M_t - B_t) h(m_t, M_t) \right] dB_t + \text{f.v.}, \\ dI_4 &= - \left[\int_{M_t}^{\infty} dy \nabla \psi_{1-t}^{(1)}(m_t - B_t, y - B_t) h(m_t, y) \right] dB_t + \text{f.v.}.\end{aligned}$$

It follows that if we denote by $X_t := \mathbb{E}\left(h(m_1, M_1)|\mathcal{F}_t\right)$ for $0 \leq t < 1$, then X is a martingale and

$$dX_t = \dot{Q}_t(h) dB_t,$$

where $\dot{Q}_t(h) := \int_{\mathbb{R}^2} \dot{Q}_t(dxdy) h(x, y)$ with

$$\begin{aligned}\dot{Q}_t(dx, dy) &:= - \left[1_{(x < m_t)} 1_{(M_t < y)} \nabla \psi_{1-t}(x - B_t, y - B_t) \right] dx dy - \\ &\quad - \left[1_{(x < m_t)} \nabla \psi_{1-t}^{(2)}(x - B_t, M_t - B_t) \right] dx \otimes \delta_{M_t}(dy) - \\ &\quad - \left[\nabla \phi_{1-t}(m_t - B_t, M_t - B_t) \right] \delta_{m_t}(dx) \otimes \delta_{M_t}(dy) - \\ &\quad - \left[1_{(y > M_t)} \nabla \psi_{1-t}^{(1)}(m_t - B_t, y - B_t) \right] \delta_{m_t}(dx) \otimes dy.\end{aligned}$$

On the other hand, we may re-write $X_t = \int_{\mathbb{R}^2} Q_t(dxdy)h(x, y)$ with

$$\begin{aligned} Q_t(dx, dy) &:= \left[1_{(x < m_t)} 1_{(M_t < y)} \psi_{1-t}(x - B_t, y - B_t) \right] dx dy + \\ &\quad + \left[1_{(x < m_t)} \psi_{1-t}^{(2)}(x - B_t, M_t - B_t) \right] dx \otimes \delta_{M_t}(dy) + \\ &\quad + \left[\phi_{1-t}(m_t - B_t, M_t - B_t) \right] \delta_{m_t}(dx) \otimes \delta_{M_t}(dy) + \\ &\quad + \left[1_{(y > M_t)} \psi_{1-t}^{(1)}(m_t - B_t, y - B_t) \right] \delta_{m_t}(dx) \otimes dy. \end{aligned}$$

Then we have $\dot{Q}_t(dxdy) = \varrho_t(x, y) Q_t(dxdy)$ with the density $\varrho_t(x, y)$ given by

$$\begin{aligned} \varrho_t(x, y) &:= -1_{(x < m_t)} 1_{(M_t < y)} \nabla \log \psi_{1-t}(x - B_t, y - B_t) \\ &\quad - 1_{(x < m_t, y = M_t)} \nabla \log \psi_{1-t}^{(2)}(x - B_t, M_t - B_t) \\ &\quad - 1_{(x = m_t, y = M_t)} \nabla \log \phi_{1-t}(m_t - B_t, M_t - B_t) \\ &\quad - 1_{(x = m_t, y > M_t)} \nabla \log \psi_{1-t}^{(1)}(m_t - B_t, y - B_t). \end{aligned} \quad (2.10)$$

The theory of enlargement of filtrations says that

$$B_t = \tilde{B}_t + \int_0^t \varrho_s(m_1, M_1) ds, \quad 0 \leq t < 1, \quad (2.11)$$

with \tilde{B} a Brownian motion with respect to $(\tilde{\mathcal{F}}_t)$.

By Cauchy-Schwarz' inequality and the L^2 -isometry for \tilde{B} , we get that

$$\mathbb{E} \left| \int_0^1 g_{m_1, M_1}(B_s) d\tilde{B}_s \right| \leq \sqrt{\mathbb{E} \int_0^1 g_{m_1, M_1}^2(B_s) ds} \leq \|g\|_\infty,$$

which in view of (2.11) imply that

$$\mathbb{E} \left| \int_0^1 g_{m_1, M_1}(B_s) dB_s \right| \leq \|g\|_\infty \left(1 + \mathbb{E} \left(\int_0^1 |\varrho_s(m_1, M_1)| ds \right) \right).$$

Then (2.5) follows once we have checked that

$$\mathbb{E} \left(\int_0^1 |\varrho_s(m_1, M_1)| ds \right) < \infty. \quad (2.12)$$

To prove (2.12), we use (2.10) and write

$$\varrho_t(m_1, M_1) = J_1(t) + J_2(t) + J_3(t) + J_4(t),$$

with obvious definitions of $J_i, 1 \leq i \leq 4$. Notice that $\{m_1 < m_t\} = \{\hat{m}_{1-t} < m_t - B_t\}$ and $\{M_t < M_1\} = \{\hat{M}_{1-t} > M_t - B_t\}$ where \hat{m}, \hat{M} are the minimum and maximum processes of $\hat{B}_s := B_{s+t} - B_s, s \geq 0$. Recalling that $\psi_{1-t}(a, b)$ are the densities of $(\hat{m}_{1-t}, \hat{M}_{1-t})$, we have that

$$\begin{aligned} \mathbb{E} [|J_1(t)|] &= \mathbb{E} \left[1_{(m_1 < m_t)} 1_{(M_t < M_1)} \left| \nabla \log \psi_{1-t}(m_1 - B_t, M_1 - B_t) \right| \right] \\ &= \mathbb{E} \int_{-\infty}^{m_t - B_t} da \int_{M_t - B_t} db \left| \nabla \psi_{1-t}(a, b) \right|. \end{aligned}$$

Elementary computations show that for any $r > 0$,

$$\nabla \psi_r(a, b) = 2 \sum_{k \in \mathbb{Z}} (-1)^k k(k+1) p_r''(b + k(b-a)),$$

where $p_r''(x) = (-\frac{1}{r} + \frac{x^2}{r^2})p_r(x)$ is the second order derivative of $p_r(\cdot)$. Bounding all terms by its absolute values, we get that

$$\mathbb{E}[|J_1(t)|] \leq 2 \mathbb{E} \int_{-\infty}^{m_t - B_t} da \int_{M_t - B_t}^{\infty} db \sum_{k \in \mathbb{Z}} |k(k+1)| \frac{1 + (b + k(b-a))^2}{(1-t)^2} p_{1-t}(b + k(b-a)).$$

In the above sum $\sum_{k \in \mathbb{Z}}$, we decompose as $\sum_{k \geq 1}$ and $\sum_{k \leq -2}$. By changing B to $-B$, it is easy to see that the two sums $\sum_{k \geq 1}$ and $\sum_{k \leq -2}$ give the same contribution in $\mathbb{E}[\int da \int db]$. It follows that

$$\begin{aligned} \mathbb{E}[|J_1(t)|] &\leq 4 \mathbb{E} \int_{-\infty}^{m_t - B_t} da \int_{M_t - B_t}^{\infty} db \sum_{k \geq 1} k(k+1) \frac{1 + (b + k(b-a))^2}{(1-t)^2} p_{1-t}(b + k(b-a)) \\ &= \frac{4}{(1-t)^2} \int_0^{\infty} dx (1+x^2) p_{1-t}(x) \mathbb{E} \left[\sum_{k=1}^{\infty} k \int_{-\infty}^{m_t - B_t} da 1_{(ka \geq (k+1)(M_t - B_t) - x)} \right] \\ &\leq \frac{4}{(1-t)^2} \int_0^{\infty} dx (1+x^2) p_{1-t}(x) \mathbb{E} \left[1_{(x \geq M_t - m_t)} \frac{x^2}{M_t - m_t} \right], \end{aligned}$$

where the above equality follows from a change of variable $x = b + k(b-a)$. If we denote by γ a Brownian motion independent of B , then for any $0 < t < 1$,

$$\mathbb{E}[|J_1(t)|] \leq \frac{4}{(1-t)^2} \mathbb{E} \left[1_{(\gamma_{1-t} \geq M_t - m_t)} \frac{(1 + \gamma_{1-t}^2) \gamma_{1-t}^2}{M_t - m_t} \right],$$

which is continuous for $t \in (0, 1)$, behaves as $\frac{c_3}{\sqrt{t}}$ as $t \rightarrow 0$ (with $c_3 := 4\mathbb{E}[1_{(\gamma_1 > 0)}(1 + \gamma_1^2)\gamma_1^2] \mathbb{E}[\frac{1}{M_1 - m_1}]$) and converges to 0 as $t \uparrow 1$ (since $\frac{1}{M_1 - m_1}$ has finite moments of any order and). Consequently,

$$\int_0^1 \mathbb{E}[|J_1(t)|] dt < \infty.$$

The other terms $J_2(t), J_3(t), J_4(t)$ can be estimated in the same way: By symmetry, $\int_0^1 \mathbb{E}[|J_4(t)|] dt = \int_0^1 \mathbb{E}[|J_2(t)|] dt$. For $J_2(t)$, we observe that for any $0 < t < 1$ and $a < 0 < b$,

$$\psi_{1-t}^{(2)}(a, b) da = \mathbb{P}(m_{1-t} \in da, M_{1-t} \leq b).$$

It follows that

$$\begin{aligned} \mathbb{E}[|J_2(t)|] &= \mathbb{E} \left[1_{(m_1 < m_t)} 1_{(M_t = M_1)} |\nabla \log \psi_{1-t}^{(2)}(m_1 - B_t, M_t - B_t)| \right] \\ &= \mathbb{E} \left[1_{(\widehat{m}_{1-t} < m_t - B_t)} 1_{(\widehat{M}_{1-t} \leq M_t - B_t)} |\nabla \log \psi_{1-t}^{(2)}(\widehat{m}_{1-t}, M_t - B_t)| \right] \\ &= \mathbb{E} \int_{-\infty}^{m_t - B_t} da |\nabla \psi_{1-t}^{(2)}(a, M_t - B_t)|. \end{aligned} \tag{2.13}$$

By (2.9) and (2.8), $\frac{\partial \psi_{1-t}^{(2)}}{\partial a}(a, b) = \int_0^b \frac{\partial \psi_{1-t}}{\partial a}(a, y) dy = 2 \sum_{k \in \mathbb{Z}} (-1)^{k+1} k^2 (p'_{1-t}(b + k(b-a)) - p'_{1-t}(-ka)) = 2 \sum_{k \in \mathbb{Z}} (-1)^{k+1} k^2 p'_{1-t}(b + k(b-a))$, by the asymmetry of $p'_{1-t}(\cdot)$. Hence,

$$\nabla \psi_{1-t}^{(2)}(a, b) = \frac{\partial \psi_{1-t}^{(2)}}{\partial a}(a, b) + \frac{\partial \psi_{1-t}^{(2)}}{\partial b}(a, b) = 2 \sum_{k \in \mathbb{Z}} (-1)^k k p'_{1-t}(b + k(b-a)).$$

It follows that

$$\begin{aligned} \int_{-\infty}^{m_t - B_t} da |\nabla \psi_{1-t}^{(2)}(a, b)| &\leq 2 \sum_{k \in \mathbb{Z}} \int_{-\infty}^{m_t - B_t} da |k p'_{1-t}(b + k(b - a))| \\ &= 2 \sum_{k \in \mathbb{Z}, k \neq 0} p_{1-t}(b + k(b - (m_t - B_t))). \end{aligned}$$

Using the elementary fact: for any $x > 0$ and $r > 0$,

$$\sum_{k \in \mathbb{Z}, k \neq 0} p_r(kx) \leq 2 \int_0^\infty p_r(tx) dt = \frac{1}{x}, \quad (2.14)$$

we deduce from (2.13) that for any $0 < t < 1$,

$$\mathbb{E}[|J_2(t)|] \leq \mathbb{E}\left(\frac{1}{M_t - m_t}\right) = \frac{1}{\sqrt{t}} \mathbb{E}\left(\frac{1}{M_1 - m_1}\right).$$

Hence

$$\int_0^1 \mathbb{E}[|J_2(t)|] dt < \infty.$$

Finally for $J_3(t)$, we have that

$$\begin{aligned} \mathbb{E}[|J_3(t)|] &= \mathbb{E}\left[1_{(m_1=m_t)} 1_{(M_t=M_1)} |\nabla \log \phi_{1-t}(m_t - B_t, M_t - B_t)|\right] \\ &= \mathbb{E}\left[1_{(\widehat{m}_{1-t} \geq m_t - B_t)} 1_{(\widehat{M}_{1-t} \leq M_t - B_t)} |\nabla \log \phi_{1-t}(m_t - B_t, M_t - B_t)|\right] \\ &= \mathbb{E}\left[|\nabla \phi_{1-t}(m_t - B_t, M_t - B_t)|\right]. \end{aligned} \quad (2.15)$$

Recalling (2.7). Elementary computations yield that

$$\nabla \phi_{1-t}(a, b) = \sum_{k \in \mathbb{Z}} (-1)^k \left[p_{1-t}(b + k(b - a)) - p_{1-t}(a + k(b - a)) \right] = 2 \sum_{k \in \mathbb{Z}} (-1)^k p_{1-t}(b + k(b - a)).$$

Hence $|\nabla \phi_{1-t}(a, b)| \leq 2p_{1-t}(a) + 2p_{1-t}(b) + 2 \sum_{j \in \mathbb{Z}, j \neq 0} p_{1-t}(j(b - a)) \leq \sqrt{\frac{8}{\pi(1-t)}} + \frac{2}{b-a}$. It follows that

$$\mathbb{E}[|J_3(t)|] \leq \sqrt{\frac{8}{\pi(1-t)}} + 2\mathbb{E}\left(\frac{1}{M_t - m_t}\right) \leq \sqrt{\frac{8}{\pi(1-t)}} + \frac{2}{\sqrt{t}} \mathbb{E}\left(\frac{1}{M_1 - m_1}\right).$$

Therefore $\int_0^1 \mathbb{E}[|J_3(t)|] dt < \infty$, which completes the proof of (2.12). Then Lemma 4 follows. \square

Remark 5. Recall that $\psi_r(a, b)$ is the density of (m_r, M_r) (cf. (2.8)). To check that $\psi_r(a, b) > 0$, M. Lifshits suggested to use the following integration equation: For any $a < 0 < b$ and any $r > s > 0$, by the Markov property at s ,

$$\begin{aligned} \psi_r(a, b) &= \int_a^b \mathbb{P}(B_s \in dx, a < m_s < M_s < b) \psi_{r-s}(a - x, b - x) \\ &= \int_a^b \Upsilon_{s,a,b}(x) \psi_{r-s}(a - x, b - x) dx, \end{aligned} \quad (2.16)$$

by using the notation in (2.6). Elementary computations show that for any $s > 0$ and $a < 0 < b$,

$$\Upsilon_{s,a,b}(x) \geq p_s(x) - p_s(2b - x) - p_s(x - 2a), \quad \forall a < x < b,$$

which implies that for any $a < 0 < b$ and $0 < s < a^2 \wedge b^2$,

$$\Upsilon_{s,a,b}(0) \geq (1 - 2e^{-2})p_s(0) > 0. \quad (2.17)$$

If $\psi_r(a, b) = 0$, then by the continuity of $\Upsilon_{s,a,b}(\cdot)$ and $\psi_{r-s}(\cdot, \cdot)$, we deduce from (2.16) that $\psi_{r-s}(a, b) = 0$ for any $0 < s < a^2 \wedge b^2 \wedge r$. By iterating this procedure, we would get that $\psi_v(a, b) = 0$ for all $0 < v < r$; but this contradicts with the fact that for any fixed $a < 0 < b$,

$$\psi_v(a, b) \sim 4 \frac{2b-a}{v} p_v(2b-a) + 4 \frac{b-2a}{v} p_v(b-2a), \quad v \rightarrow 0,$$

which is positive for all small v . Hence $\psi_r(a, b) > 0$ for any $r > 0$ and $a < 0 < b$.

Remark 6. If we enlarge the filtration only with M_1 , it was known that (see Jeulin and Yor [8])

$$B_t = \widehat{B}_t + \int_0^t \frac{ds}{\sqrt{1-s}} \left(1_{(M_s < M_1)} \frac{M_1 - B_s}{\sqrt{1-s}} - 1_{(M_s = M_1)} k\left(\frac{M_1 - B_s}{\sqrt{1-s}}\right) \right), \quad 0 \leq t < 1,$$

with \widehat{B} a Brownian motion in the filtration $(\mathcal{F}_t \vee \sigma\{M_1\})_{0 \leq t \leq 1}$ and $k(x) := e^{-\frac{x^2}{2}} \frac{1}{\int_0^x d r e^{-r^2/2}}$.

Let

$$H_\lambda := \int W_\lambda(y) dL_1(y), \quad \text{with } W_\lambda(y) = \frac{W(\lambda y)}{(2\lambda \log \log \lambda)^{1/2}}, \quad y \in \mathbb{R}.$$

By using (2.4), it is easy to check that \mathbb{P}_W -almost surely, $H_\lambda \in L^1(B) \equiv L^1(\mathbb{P}_B, \sigma\{B_s, 0 \leq s \leq 1\})$. Write $d_{L^1}(\xi, \eta)$ for the distance in $L^1(B)$ for any $\xi, \eta \in L^1(B)$.

Lemma 7. $\mathbb{P}(dW)$ -almost surely,

$$d_{L^1}(H_\lambda, \mathcal{K}_B) \rightarrow 0, \quad \text{as } \lambda \rightarrow \infty,$$

where \mathcal{K}_B is defined in (1.2). Moreover, \mathbb{P}_W -almost surely for any $\xi \in \mathcal{K}_B$, $\liminf_{\lambda \rightarrow \infty} d(H_\lambda, \xi) = 0$.

It easy to see that $\mathcal{K}_B \subset L^1(B)$, since by Cauchy-Schwarz' inequality, $(\int_{m_1}^{M_1} f(x) dL_1(x))^2 \leq (\int L_1(x)^2 dx) (\int_{m_1}^{M_1} (f(x))^2 dx) \leq \sup_x L_1(x) \in L^p(B)$ for any $p > 0$.

Proof: Let $\varepsilon > 0$. Applying Theorem 3 to $s = m_1$ and $r = M_1$, we get that \mathbb{P}_W -a.s. for all large $\lambda \geq \lambda_0(W)$, there exists a (random) function $f \equiv f_\lambda \in \mathcal{K}_{m_1, M_1}$ such that $\sup_{m_1 \leq x \leq M_1} |W_\lambda(x) - f(x)| \leq \varepsilon$. Let $g(x) := W_\lambda(x) - f(x)$. Notice that $\sup_{m_1 \leq x \leq M_1} |g(x)| \leq \varepsilon$, and conditioned on W , g is measurable with respect to $\sigma\{m_1, M_1\}$. By Lemma 4, $\left\| \int g(x) dL_1(x) \right\|_1 \leq c_1 \varepsilon$. Hence $d_{L^1}(H_\lambda, \mathcal{K}_B) \rightarrow 0$. The another part of the Lemma is again a consequence of Strassen's law of iterated logarithm (Theorem 3) and Lemma 4. \square

We now are ready to give the proof of Theorems 2 and 1.

Proof of Theorem 2. Firstly, we remark that by Brownian scaling, \mathbb{P}_W -a.s.,

$$\frac{Z_t}{t^{3/4}} \stackrel{(d)}{=} - \int_{m_1}^{M_1} \frac{1}{t^{1/4}} W(\sqrt{t}y) dL_1(y). \quad (2.18)$$

In fact, by the change of variables $x = y\sqrt{t}$, we get

$$\int_{-\infty}^{+\infty} L_t(x) dW(x) = \sqrt{t} \int_{-\infty}^{+\infty} \left(\frac{L_t(y\sqrt{t})}{\sqrt{t}} \right) dW(y\sqrt{t})$$

which has the same distribution as

$$\sqrt{t} \int_{-\infty}^{+\infty} L_1(y) dW(y\sqrt{t})$$

from the scaling property of the local time of the Brownian motion. Since $(L_1(x))_{x \in \mathbb{R}}$ is a continuous semi-martingale, independent from the process W , from the formula of integration by parts, we get that \mathbb{P}_W -a.s.,

$$\sqrt{t} \int_{-\infty}^{+\infty} L_1(y) dW(y\sqrt{t}) = -t^{3/4} \int_{m_1}^{M_1} \left(\frac{W(\sqrt{t}y)}{t^{1/4}} \right) dL_1(y),$$

yielding (2.18). Theorem 2 follows from Lemma 7. \square

Proof of Theorem 1. We use the strong approximation of Zhang [15] : there exists on a suitably enlarged probability space, a coupling of ξ , S , B and W such that (ξ, W) is independent of (S, B) and for any $\varepsilon > 0$, almost surely,

$$\max_{0 \leq m \leq n} |K(m) - Z(m)| = o(n^{\frac{1}{2} + \frac{1}{2(2+\delta)} + \varepsilon}), \quad n \rightarrow +\infty. \quad (2.19)$$

From the independence of (ξ, W) and (S, B) , we deduce that for \mathbb{P} -a.e. (ξ, W) , under the quenched probability $\mathbb{P}(\cdot | \xi, W)$, the limit points of the laws of \tilde{K}_n and \tilde{Z}_n are the same ones. Now, by adapting the proof of Theorem 2, we have that for \mathbb{P} -a.e. (ξ, W) , under the quenched probability $\mathbb{P}(\cdot | \xi, W)$, the limit points of the laws of \tilde{Z}_n , as $n \rightarrow \infty$, under the topology of weak convergence of measures, are equal to the closure of the laws of random variables in \mathcal{K}_B . It gives that for \mathbb{P} -a.e. (ξ, W) , under the quenched probability $\mathbb{P}(\cdot | \xi, W)$, the limit points of the laws of \tilde{K}_n , as $n \rightarrow \infty$, under the topology of weak convergence of measures, are equal to the closure of the laws of random variables in \mathcal{K}_B and Theorem 1 follows. \square

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