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## The maximal drawdown of the Brownian meander

by

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**Summary.** Motivated by evaluating the limiting distribution of randomly biased random walks on trees, we compute the exact value of a negative moment of the maximal drawdown of the standard Brownian meander.

**Keywords.** Brownian meander, Bessel process, maximal drawdown.

**2010** Mathematics Subject Classification. 60J65.

### 1 Introduction

Let  $(X(t), t \in [0, 1])$  be a random process. Its maximal drawdown on [0, 1] is defined by

$$X^{\#}(1) := \sup_{s \in [0, 1]} \left[ \overline{X}(s) - X(s) \right],$$

where  $\overline{X}(s) := \sup_{u \in [0,s]} X(u)$ . There has been some recent research interest on the study of drawdowns from probabilistic point of view ([7], [8]) as well as applications in insurance and finance ([1], [2], [3], [10], [12]).

We are interested in the maximal drawdown  $\mathfrak{m}^{\#}(1)$  of the standard Brownian meander  $(\mathfrak{m}(t), t \in [0, 1])$ . Our motivation is the presence of the law of  $\mathfrak{m}^{\#}(1)$  in the limiting distribution of randomly biased random walks on supercritical Galton–Watson trees ([4]); in particular, the value of  $\mathbb{E}(\frac{1}{\mathfrak{m}^{\#}(1)})$  is the normalizing constant in the density function of

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this limiting distribution. The sole aim of the present note is to compute  $\mathbb{E}(\frac{1}{\mathfrak{m}^{\#}(1)})$ , which turns out to have a nice numerical value.

Let us first recall the definition of the Brownian meander. Let  $W := (W(t), t \in [0, 1])$  be a standard Brownian motion, and let  $\mathfrak{g} := \sup\{t \leq 1 : W(t) = 0\}$  be the last passage time at 0 before time 1. Since  $\mathfrak{g} < 1$  a.s., we can define

$$\mathfrak{m}(s) := \frac{|W(\mathfrak{g} + s(1 - \mathfrak{g}))|}{(1 - \mathfrak{g})^{1/2}}, \qquad s \in [0, 1].$$

The law of  $(\mathfrak{m}(s), s \in [0, 1])$  is called the law of the standard Brownian meander. For an account of general properties of the Brownian meander, see Yen and Yor [11].

**Theorem 1.1.** Let  $(\mathfrak{m}(s), s \in [0, 1])$  be a standard Brownian meander. We have

(1.1) 
$$\mathbb{E}\left(\frac{1}{\sup_{s \in [0,1]} \left[\overline{\overline{\mathfrak{m}}}(s) - \mathfrak{m}(s)\right]}\right) = \left(\frac{\pi}{2}\right)^{1/2},$$

where  $\overline{\mathfrak{m}}(s) := \sup_{u \in [0, s]} \mathfrak{m}(u)$ .

The theorem is proved in Section 2.

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N.B. from the first-named coauthors: This note originates from a question we asked our teacher, **Professor Marc Yor (1949–2014)**, who passed away in January 2014, during the preparation of this note. He provided us, in November 2012, with the essential of the material in Section 2.

# 2 Proof

Let  $R := (R(t), t \ge 0)$  be a three-dimensional Bessel process with R(0) = 0, i.e., the Euclidean modulus of a standard three-dimensional Brownian motion. The proof of Theorem 1.1 relies on an absolute continuity relation between  $(\mathfrak{m}(s), s \in [0, 1])$  and  $(R(s), s \in [0, 1])$ , recalled as follows.

Fact 2.1. (Imhof [5]) Let  $(\mathfrak{m}(s), s \in [0, 1])$  be a standard Brownian meander. Let  $(R(s), s \in [0, 1])$  be a three-dimensional Bessel process with R(0) = 0. For any measurable and non-negative functional F, we have

$$\mathbb{E}\Big[F(\mathfrak{m}(s),\, s \in [0,\,1])\Big] = \left(\frac{\pi}{2}\right)^{1/2} \mathbb{E}\Big[\frac{1}{R(1)}\, F(R(s),\, s \in [0,\,1])\Big]\,.$$

We now proceed to the proof of Theorem 1.1. Let

$$L := \mathbb{E}\left(\frac{1}{\sup_{s \in [0,1]} \left[\overline{\mathfrak{m}}(s) - \mathfrak{m}(s)\right]}\right).$$

Write  $\overline{R}(t) := \sup_{u \in [0, t]} R(u)$  for  $t \ge 0$ . By Fact 2.1,

$$L = \left(\frac{\pi}{2}\right)^{1/2} \mathbb{E}\left[\frac{1}{R(1)} \frac{1}{\sup_{s \in [0,1]} \left[\overline{R}(s) - R(s)\right]}\right]$$
$$= \left(\frac{\pi}{2}\right)^{1/2} \int_{0}^{\infty} \mathbb{E}\left[\frac{1}{R(1)} \mathbf{1}_{\left\{\sup_{s \in [0,1]} \left[\overline{R}(s) - R(s)\right] < \frac{1}{a}\right\}}\right] da,$$

the last equality following from the Fubini–Tonelli theorem. By the scaling property,  $\mathbb{E}\left[\frac{1}{R(1)}\,\mathbf{1}_{\{\sup_{s\in[0,\,1]}[\overline{R}(s)-R(s)]<\frac{1}{a}\}}\right] = \mathbb{E}\left[\frac{a}{R(a^2)}\,\mathbf{1}_{\{\sup_{u\in[0,\,a^2]}[\overline{R}(u)-R(u)]<1\}}\right] \text{ for all } a>0. \text{ So by means of a change of variables } b=a^2, \text{ we obtain:}$ 

$$L = \left(\frac{\pi}{8}\right)^{1/2} \int_0^\infty \mathbb{E}\left[\frac{1}{R(b)} \mathbf{1}_{\{\sup_{u \in [0, b]} [\overline{R}(u) - R(u)] < 1\}}\right] db.$$

Define, for any random process X,

$$\tau_1^X := \inf\{t \ge 0 : \overline{X}(t) - X(t) \ge 1\},\,$$

with  $\overline{X}(t) := \sup_{s \in [0, t]} X(s)$ . For any b > 0, the event  $\{\sup_{u \in [0, b]} [\overline{R}(u) - R(u)] < 1\}$  means  $\{\tau_1^R > b\}$ , so

$$L = \left(\frac{\pi}{8}\right)^{1/2} \int_0^\infty \mathbb{E}\left[\frac{1}{R(b)} \, \mathbf{1}_{\{\tau_1^R > b\}}\right] db = \left(\frac{\pi}{8}\right)^{1/2} \mathbb{E}\left(\int_0^{\tau_1^R} \frac{1}{R(b)} \, db\right),$$

the second identity following from the Fubini–Tonelli theorem. According to a relation between Bessel processes of dimensions three and four (Revuz and Yor [9], Proposition XI.1.11, applied to the parameters p = q = 2 and  $\nu = \frac{1}{2}$ ),

$$R(t) = U\left(\frac{1}{4} \int_0^t \frac{1}{R(b)} db\right), \qquad t \ge 0,$$

where  $U := (U(s), s \ge 0)$  is a four-dimensional squared Bessel process with U(0) = 0; in other words, U is the square of the Euclidean modulus of a standard four-dimensional Brownian motion.

Let us introduce the increasing functional  $\sigma(t) := \frac{1}{4} \int_0^t \frac{1}{R(b)} db$ ,  $t \ge 0$ . We have  $R = U \circ \sigma$ , and

$$\begin{split} \tau_1^R &= \inf\{t \geq 0 : \overline{R}(t) - R(t) \geq 1\} \\ &= \inf\{t \geq 0 : \overline{U}(\sigma(t)) - U(\sigma(t)) \geq 1\} \\ &= \inf\{\sigma^{-1}(s) : s \geq 0 \text{ and } \overline{U}(s) - U(s) \geq 1\} \end{split}$$

which is  $\sigma^{-1}(\tau_1^U)$ . So  $\tau_1^U = \sigma(\tau_1^R)$ , i.e.,

$$\int_0^{\tau_1^R} \frac{1}{R(b)} \, \mathrm{d}b = 4\tau_1^U \,,$$

which implies that

$$L = (2\pi)^{1/2} \mathbb{E}(\tau_1^U)$$
.

The Laplace transform of  $\tau_1^U$  is determined by Lehoczky [6], from which, however, it does not seem obvious to deduce the value of  $\mathbb{E}(\tau_1^U)$ . Instead of using Lehoczky's result directly, we rather apply his method to compute  $\mathbb{E}(\tau_1^U)$ . By Itô's formula,  $(U(t)-4t, t \geq 0)$  is a continuous martingale, with quadratic variation  $4\int_0^t U(s) \, \mathrm{d}s$ ; so applying the Dambis-Dubins-Schwarz theorem (Revuz and Yor [9], Theorem V.1.6) to  $(U(t)-4t, t \geq 0)$  yields the existence of a standard Brownian motion  $B = (B(t), t \geq 0)$  such that

$$U(t) = 2B(\int_0^t U(s) ds) + 4t, \qquad t \ge 0.$$

Taking  $t := \tau_1^U$ , we get

$$U(\tau_1^U) = 2B(\int_0^{\tau_1^U} U(s) \, ds) + 4\tau_1^U.$$

We claim that

(2.1) 
$$\mathbb{E}\left[B\left(\int_{0}^{\tau_{1}^{U}}U(s)\,\mathrm{d}s\right)\right]=0.$$

Then  $\mathbb{E}(\tau_1^U) = \frac{1}{4} \mathbb{E}[U(\tau_1^U)]$ ; hence

(2.2) 
$$L = (2\pi)^{1/2} \mathbb{E}(\tau_1^U) = (\frac{\pi}{8})^{1/2} \mathbb{E}[U(\tau_1^U)].$$

Let us admit (2.1) for the moment, and prove the theorem by computing  $\mathbb{E}[U(\tau_1^U)]$  using Lehoczky [6]'s method; in fact, we determine the law of  $U(\tau_1^U)$ .

**Lemma 2.2.** The law of  $U(\tau_1^U)$  is given by

$$\mathbb{P}\{U(\tau_1^U) > a\} = (a+1)e^{-a}, \quad \forall a > 0.$$

In particular,

$$\mathbb{E}[U(\tau_1^U)] = \int_0^\infty (a+1)e^{-a} da = 2.$$

Since  $L = (\frac{\pi}{8})^{1/2} \mathbb{E}[U(\tau_1^U)]$  (see (2.2)), this yields  $L = (\frac{\pi}{2})^{1/2}$  as stated in Theorem 1.1.

The rest of the note is devoted to the proof of Lemma 2.2 and (2.1).

Proof of Lemma 2.2. Fix b > 1. We compute the probability  $\mathbb{P}\{\overline{U}(\tau_1^U) > b\}$  which, due to the equality  $\overline{U}(\tau_1^U) = U(\tau_1^U) + 1$ , coincides with  $\mathbb{P}\{U(\tau_1^U) > b - 1\}$ . By applying the strong Markov property at time  $\sigma_0^U := \inf\{t \geq 0 : U(t) = 1\}$ , we see that the value of  $\mathbb{P}\{\overline{U}(\tau_1^U) > b\}$  does not change if the squared Bessel process U starts at U(0) = 1. Indeed, observing that  $\sigma_0^U \leq \tau_1^U$ ,  $U(\sigma_0^U) = 1$  and that  $\overline{U}(\tau_1^U) = \sup_{s \in [\sigma_0^U, \tau_1^U]} U(s)$ , we have

$$\mathbb{P}\{\overline{U}(\tau_1^U) > b\} = \mathbb{P}\Big\{\sup_{s \in [\sigma_0^U, \tau_1^U]} U(s) > b\Big\} = \mathbb{P}_1\{\overline{U}(\tau_1^U) > b\},$$

the subscript 1 in  $\mathbb{P}_1$  indicating the initial value of U. More generally, for  $x \geq 0$ , we write  $\mathbb{P}_x(\bullet) := \mathbb{P}(\bullet \mid U(0) = x)$ ; so  $\mathbb{P} = \mathbb{P}_0$ .

Let  $b_0 = 1 < b_1 < \dots < b_n := b$  be a subdivision of [1, b] such that  $\max_{1 \le i \le n} (b_i - b_{i-1}) \to 0$ ,  $n \to \infty$ . Consider the event  $\{\overline{U}(\tau_1^U) > b\}$ : since U(0) = 1, this means U hits position b before time  $\tau_1^U$ ; for all  $i \in [1, n-1] \cap \mathbb{Z}$ , starting from position  $b_i$ , U must hit  $b_{i+1}$  before hitting  $b_i - 1$  (caution: not to be confused with  $b_{i-1}$ ). More precisely, let  $\sigma_i^U := \inf\{t \ge 0 : U(t) = b_i\}$  and let  $U_i(s) := U(s + \sigma_i^U)$ ,  $s \ge 0$ ; then

$$\{\overline{U}(\tau_1^U) > b\} \subset \bigcap_{i=1}^{n-1} \{U_i \text{ hits } b_{i+1} \text{ before hitting } b_i - 1\}.$$

By the strong Markov property, the events  $\{U_i \text{ hits } b_{i+1} \text{ before hitting } b_i - 1\}$ ,  $1 \leq i \leq n-1$ , are independent (caution: the processes  $(U_i(s), s \geq 0)$ ,  $1 \leq i \leq n-1$ , are not independent). Hence

(2.3) 
$$\mathbb{P}_1\{\overline{U}(\tau_1^U) > b\} \le \prod_{i=1}^{n-1} \mathbb{P}_{b_i}\{U \text{ hits } b_{i+1} \text{ before hitting } b_i - 1\}.$$

Conversely, let  $\varepsilon > 0$ , and if  $\max_{1 \le i \le n} (b_i - b_{i-1}) < \varepsilon$ , then we also have

$$\mathbb{P}_1\{\overline{U}(\tau_{1+\varepsilon}^U) > b\} \ge \prod_{i=1}^{n-1} \mathbb{P}_{b_i}\{U \text{ hits } b_{i+1} \text{ before hitting } b_i - 1\},$$

with  $\tau_{1+\varepsilon}^U := \inf\{t \geq 0 : \overline{U}(t) - U(t) \geq 1 + \varepsilon\}$ . By scaling,  $\overline{U}(\tau_{1+\varepsilon}^U)$  has the same distribution as  $(1+\varepsilon)\overline{U}(\tau_1^U)$ . So, as long as  $\max_{1\leq i\leq n}(b_i-b_{i-1})<\varepsilon$ , we have

$$\mathbb{P}_1\{\overline{U}(\tau_1^U) > b\} \leq \prod_{i=1}^{n-1} \mathbb{P}_{b_i}\{U \text{ hits } b_{i+1} \text{ before hitting } b_i - 1\} \leq \mathbb{P}_1\{\overline{U}(\tau_1^U) > \frac{b}{1+\varepsilon}\}.$$

Since  $\frac{1}{x}$  is a scale function for U, we have

$$\mathbb{P}_{b_i}\{U \text{ hits } b_{i+1} \text{ before hitting } b_i - 1\} = \frac{\frac{1}{b_i - 1} - \frac{1}{b_i}}{\frac{1}{b_i - 1} - \frac{1}{b_{i+1}}} = 1 - \frac{\frac{1}{b_i} - \frac{1}{b_{i+1}}}{\frac{1}{b_i - 1} - \frac{1}{b_{i+1}}}.$$

If  $\lim_{n\to\infty} \max_{0\leq i\leq n-1} (b_{i+1}-b_i)=0$ , then for  $n\to\infty$ ,

$$\sum_{i=1}^{n-1} \frac{\frac{1}{b_i} - \frac{1}{b_{i+1}}}{\frac{1}{b_i - 1} - \frac{1}{b_{i+1}}} = \sum_{i=1}^{n-1} \frac{b_i - 1}{b_i} (b_{i+1} - b_i) + o(1)$$

$$\rightarrow \int_1^b \frac{r - 1}{r} dr$$

$$= b - 1 - \log b.$$

Therefore,

$$\lim_{n \to \infty} \prod_{i=1}^{n-1} \mathbb{P}_{b_i} \{ U \text{ hits } b_{i+1} \text{ before hitting } b_i - 1 \} = e^{-(b-1-\log b)} = b e^{-(b-1)}.$$

Consequently,

$$\mathbb{P}\{\overline{U}(\tau_1^U) > b\} = b e^{-(b-1)}, \quad \forall b > 1.$$

We have already noted that  $U(\tau_1^U) = \overline{U}(\tau_1^U) - 1$ . This completes the proof of Lemma 2.2.

Proof of (2.1). The Brownian motion B being the Dambis–Dubins–Schwarz Brownian motion associated with the continuous martingale  $(U(t)-4t, t \geq 0)$ , it is a  $(\mathscr{G}_r)_{r\geq 0}$ -Brownian motion (Revuz and Yor [9], Theorem V.1.6), where, for  $r\geq 0$ ,

$$\mathscr{G}_r := \mathscr{F}_{C(r)}, \qquad C(r) := A^{-1}(r), \qquad A(t) := \int_0^t U(s) \, \mathrm{d}s,$$

and  $A^{-1}$  denotes the inverse of A. [We mention that  $\mathscr{F}_{C(r)}$  is well defined because C(r) is an  $(\mathscr{F}_t)_{t\geq 0}$ -stopping time.] As such,

$$\int_0^{\tau_1^U} U(s) \, \mathrm{d}s = A(\tau_1^U) \,.$$

For all  $r \geq 0$ ,  $\{A(\tau_1^U) > r\} = \{\tau_1^U > C(r)\} \in \mathscr{F}_{C(r)} = \mathscr{G}_r$  (observing that  $\tau_1^U$  is an  $(\mathscr{F}_t)_{t \geq 0}$ -stopping time), which means that  $A(\tau_1^U)$  is a  $(\mathscr{G}_r)_{r \geq 0}$ -stopping time. If  $A(\tau_1^U) = \int_0^{\tau_1^U} U(s) \, \mathrm{d}s$  has a finite expectation, then we are entitled to apply the (first) Wald identity to see that  $\mathbb{E}[B(A(\tau_1^U))] = 0$  as claimed in (2.1).

It remains to prove that  $\mathbb{E}[A(\tau_1^U)] < \infty$ .

Recall that U is the square of the Euclidean modulus of an  $\mathbb{R}^4$ -valued Brownian motion. By considering only the first coordinate of this Brownian motion, say  $\beta$ , we have

$$\mathbb{P}\Big\{\sup_{s\in[0,\,a]}U(s)< a^{1-\varepsilon}\Big\} \leq \mathbb{P}\Big\{\sup_{s\in[0,\,a]}|\beta(s)|< a^{(1-\varepsilon)/2}\Big\} = \mathbb{P}\Big\{\sup_{s\in[0,\,1]}|\beta(s)|< a^{-\varepsilon/2}\Big\};$$

so by the small ball probability for Brownian motion, we obtain:

$$\mathbb{P}\Big\{\sup_{s\in[0,\,a]}U(s)< a^{1-\varepsilon}\Big\} \le \exp(-c_1\,a^{\varepsilon})\,,$$

for all  $a \geq 1$  et all  $\varepsilon \in (0, 1)$ , with some constant  $c_1 = c_1(\varepsilon) > 0$ . On the event  $\{\sup_{s \in [0, a]} U(s) \geq a^{1-\varepsilon}\}$ , if  $\tau_1^U > a$ , then for all  $i \in [1, a^{1-\varepsilon} - 1] \cap \mathbb{Z}$ , the squared Bessel process U, starting from i, must first hit position i + 1 before hitting i - 1 (which, for each i, can be realized with probability  $\leq 1 - c_2$ , where  $c_2 \in (0, 1)$  is a constant that does not depend on i, nor on a). Accordingly,<sup>1</sup>

$$\mathbb{P}\Big\{\sup_{s \in [0, a]} U(s) \ge a^{1-\varepsilon}, \ \tau_1^U > a\Big\} \le (1 - c_2)^{\lfloor a^{1-\varepsilon} - 1 \rfloor} \le \exp(-c_3 a^{1-\varepsilon}),$$

with some constant  $c_3 > 0$ , uniformly in  $a \ge 2$ . We have thus proved that for all  $a \ge 2$  and all  $\varepsilon \in (0, 1)$ ,

$$\mathbb{P}\{\tau_1^U > a\} \le \exp(-c_3 a^{1-\varepsilon}) + \exp(-c_1 a^{\varepsilon}).$$

Taking  $\varepsilon := \frac{1}{2}$ , we see that there exists a constant  $c_4 > 0$  such that

$$\mathbb{P}\{\tau_1^U > a\} < \exp(-c_4 a^{1/2}), \quad \forall a > 2.$$

On the other hand, U being a squared Bessel process, we have, for all a > 0 and all  $b \ge a^2$ ,

$$\mathbb{P}\{A(a) \ge b\} = \mathbb{P}\{A(1) \ge \frac{b}{a^2}\} \le \mathbb{P}\left\{\sup_{s \in [0, 1]} U(s) \ge \frac{b}{a^2}\right\} \le e^{-c_5 b/a^2},$$

for some constant  $c_5 > 0$ . Hence, for  $b \ge a^2$  and  $a \ge 2$ ,

$$\mathbb{P}\{A(\tau_1^U) \ge b\} \le \mathbb{P}\{\tau_1^U > a\} + \mathbb{P}\{A(a) \ge b\} \le \exp(-c_4 a^{1/2}) + e^{-c_5 b/a^2}.$$

Taking  $a := b^{2/5}$  gives that

$$\mathbb{P}\{A(\tau_1^U) > b\} < \exp(-c_6 b^{1/5})$$
.

for some constant  $c_6 > 0$  and all  $b \ge 4$ . In particular,  $\mathbb{E}[A(\tau_1^U)] < \infty$  as desired.

<sup>&</sup>lt;sup>1</sup>This is the special case  $b_i := i$  of the argument we have used to obtain (2.3).

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