

Let  $(B_t)_{t \geq 0}$  be a Brownian motion on  $\mathbb{R}$ .

Consider the **running maximum**  $B_t^{\max} = \max_{s \leq t} B_s$ .

For discrete-time submartingales, we proved several

useful estimates for the values of running maxima  $B_t^* = |B|_t^{\max}$ .

We could (given more time) prove the same results for continuous-time martingales like Brownian motion. But we can prove stronger results here.

**Theorem:** (Bachelier) For each  $t \geq 0$ ,  $B_t^{\max} \stackrel{d}{=} |B_t|$ .

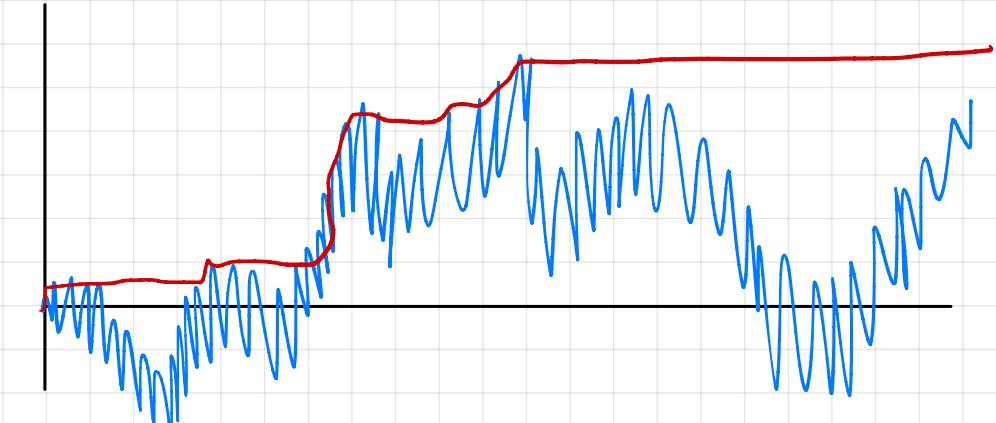
In fact, we'll compute (most of) the joint distribution of  $(B_t^{\max}, B_t)$ :

$$\begin{aligned} \overset{\circ}{P}(B_t^{\max} \geq z, B_t < z-y) &= \overset{\circ}{P}(B_t > z-y) \\ \underset{z \geq 0}{\uparrow} \quad \underset{y > 0}{\uparrow} &= P(Z > \frac{z-y}{\sqrt{t}}) \quad z \stackrel{d}{=} N(0,1) \end{aligned}$$

Note: pathwise, the processes  $(B_t^{\max})_{t \geq 0}$ ,  $(|B_t|)_{t \geq 0}$  are **very** different.

Bessel process

as rough as  $B_t$ .



Theorem: Let  $z \geq 0, y \geq 0$ .  $P(B_{t_f}^{\max} \geq z, B_t < z-y) = P(B_t > z+y)$ ,  $\therefore B_t^{\max} \stackrel{d}{=} |B_t|$ .

Pf. Let  $\tau = T_z$ , the  $B$ -Hitting time of  $\{z\}$

$\leftarrow$  closed  
 $\therefore$  optimal time.

Let  $(\tilde{B}_t)_{t \geq 0}$  be the Brownian motion reflected at time  $\tau$ ,

$$\tilde{B}_t = B_{t \wedge \tau} - (B_t - B_{t \wedge \tau})$$

$\uparrow$  Another Brownian motion, by Reflection Principle.

Then on  $\{\tau \leq t\}$ ,  $\{B_t < z-y\} = \{\tilde{B}_t > z+y\} \subseteq \{\tilde{\tau} \leq t\}$

$$\{ \max_{s \leq t} B_s \geq z \} = \{ B_{t_f}^{\max} \geq z \}$$

$$\therefore \{ B_{t_f}^{\max} \geq z, B_t < z-y \} = \{ \tau \leq t, \tilde{B}_t > z+y \}$$

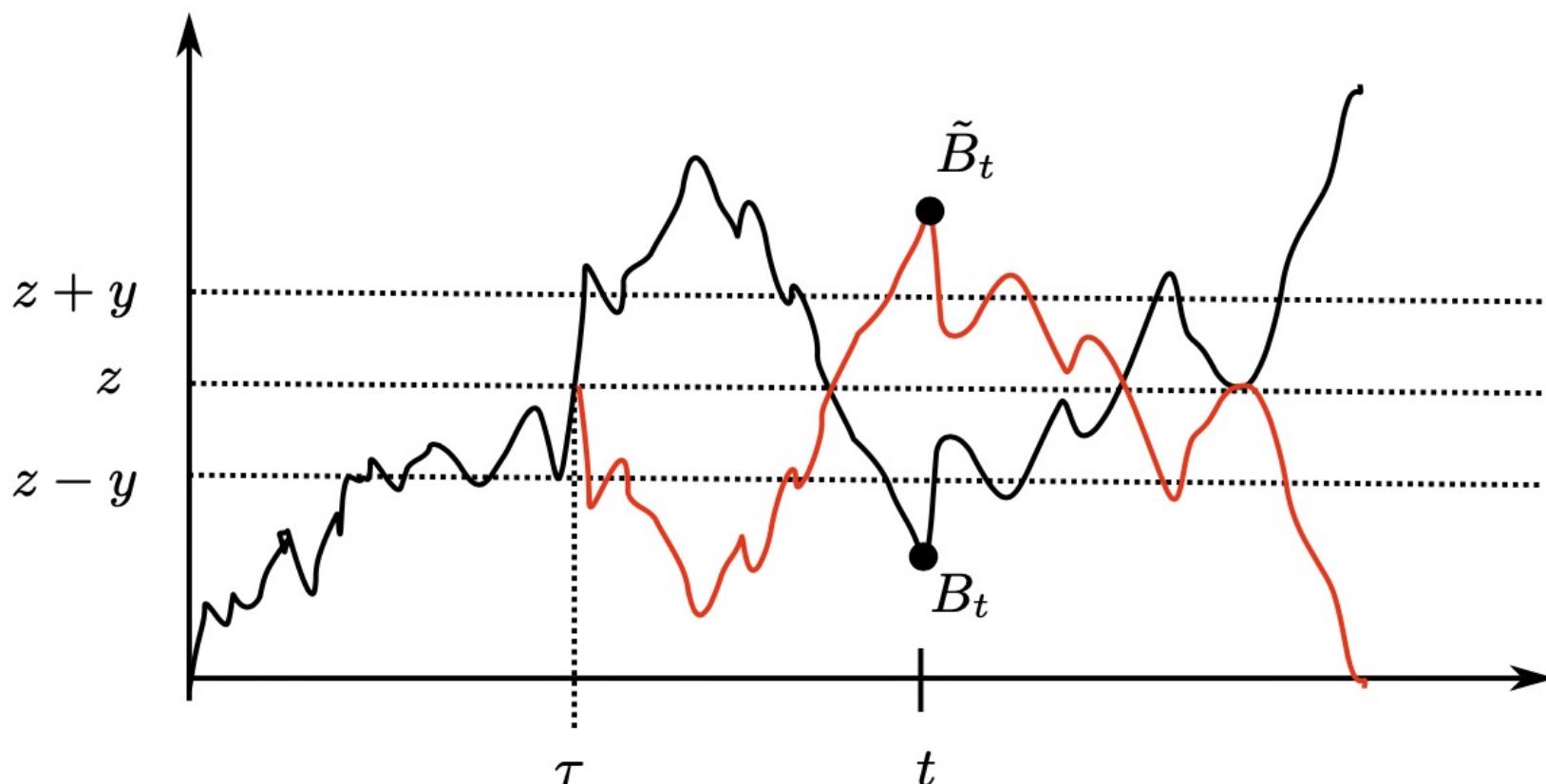
$$= \{ \tilde{\tau} \leq t, \tilde{B}_t > z+y \}$$

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Note:  $B_t = \tilde{B}_t$  for  $t \leq \tau$ ;

$$\therefore \tilde{\tau} = \inf \{ t > 0 : \tilde{B}_t = z \}$$

$$= \tau$$



Thus  $P^o(B_t^{\max} \geq z, B_t < z-y) = P^o(B_t > z+y)$  for  $z > 0, y \geq 0$ .

In particular,

$$\begin{aligned}
 P^o(B_t^{\max} \geq z) &= P^o(B_t^{\max} \geq z, B_t \geq z) + P^o(B_t^{\max} \geq z, B_t < z) \\
 &\stackrel{\swarrow}{\leq} P^o(B_t^* \geq z) \\
 &\stackrel{\text{II}}{=} P^o(B_t \geq z) + P^o(B_t > z+0) \\
 &\stackrel{\text{II}}{=} 2P^o(B_t > z) = P^o(B_t > z) + P^o(B_t < -z) \quad \text{b/c } B_t \stackrel{d}{=} -B_t \\
 &= P(|B_t| > z) \\
 &\stackrel{\text{III}}{=} 4P(|B_t| \geq z) \\
 &\leq 2e^{-z^2/2t}.
 \end{aligned}$$

Cor: The hitting time  $T_z$  of  $\{z\}$  satisfies

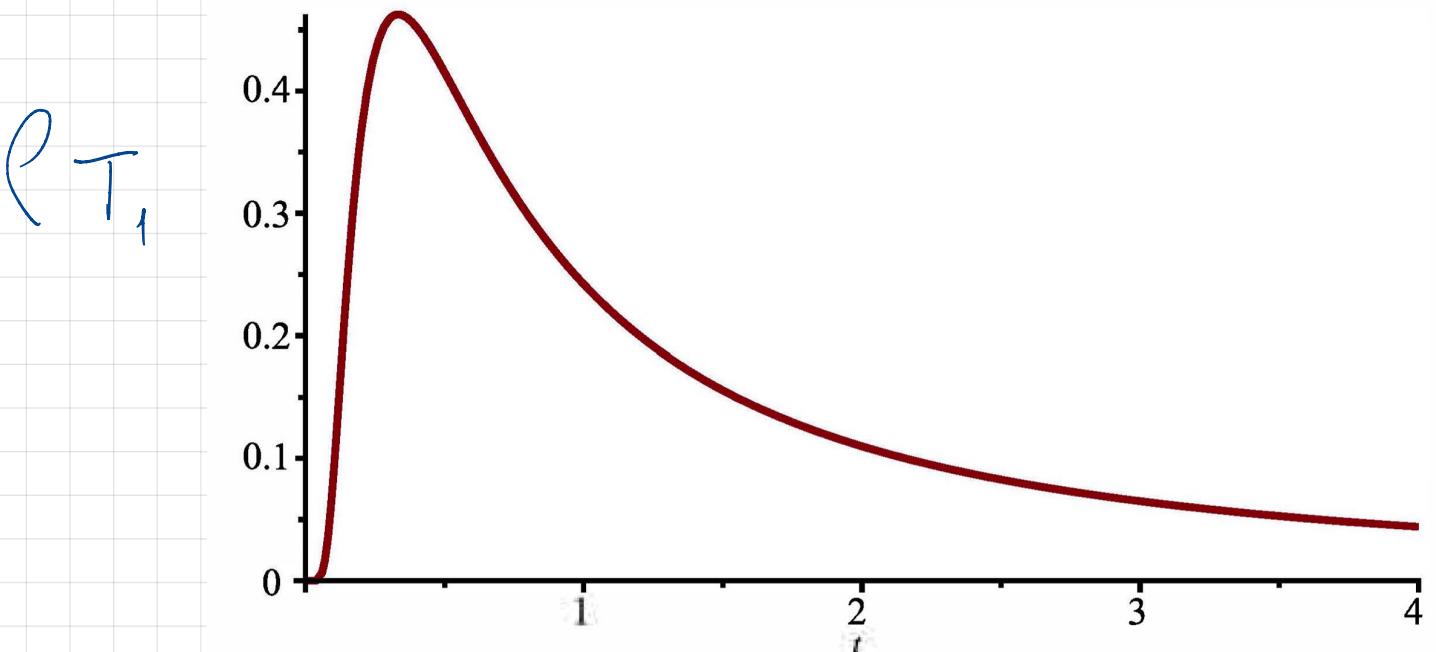
$$\begin{aligned}
 P^o(T_z \leq t) &= P^o(B_t^{\max} \geq z) \quad \text{subs.} \\
 \therefore 2P^o(B_t \geq z) &= 2 \int_z^\infty \frac{1}{\sqrt{2\pi t}} e^{-x^2/2t} dx \\
 &= 2 \int_{-\infty}^0 \frac{1}{\sqrt{2\pi t}} e^{-x^2/2t} (z\sqrt{t} - x\sqrt{t}) dx \\
 &= \int_0^t \frac{z}{\sqrt{2\pi u^3}} e^{-z^2/2u} du
 \end{aligned}$$

That is: we've explicitly calculated the probability density of  $T_z$ !

$$\rho_{T_z}(t) = \frac{z}{\sqrt{2\pi t^3}} e^{-z^2/2t} \mathbb{I}_{(0,\infty)}(t). \quad \leftarrow \alpha > 0$$

$$\rho_{T_{\alpha z}}(\alpha t) = \frac{1}{\alpha} \rho_{T_z}(t)$$

$$\therefore \rho_{T_{\alpha z}}(s) = \frac{1}{\alpha} \rho_{T_z}(s/\alpha)$$



$$\text{In particular, } P(T_z < \infty) = 1 \quad \forall z \quad T_{-\beta}^B = T_z^B = T_z$$

(Already knew this:  $\limsup_{t \rightarrow \infty} t \beta_t = \infty$  a.s.)

But

$$\mathbb{E}^0[T_z] = \int_0^\infty t \cdot \rho_{T_z}(t) dt$$

$\therefore$  I-dm  $B$ , M.  
 $\beta$  null-current!

$$= \frac{z}{\sqrt{2\pi}} \int_0^\infty \frac{1}{\sqrt{t}} e^{-z^2/2t} dt \sim \int_0^\infty \frac{1}{\sqrt{t}} dt = \infty$$

bounded in  $[0,1]$   
 $\rightarrow$  as  $t \rightarrow \infty$ .

We can use this density to compute other distributions exactly.

Prop: (Arcsine Law) Let  $L_1 = \sup\{0 \leq t \leq 1 : B_t = 0\}$ .

$$L_1 \stackrel{d}{=} \sin^2 U,$$

$$U \stackrel{d}{=} \text{Unif}[0,1]$$

$$\text{Then } P(L_1 \leq t) = \frac{2}{\pi} \arcsin(\sqrt{t}), \quad 0 \leq t \leq 1.$$

$$P_{L_1}(r) = \frac{1}{\pi} \frac{1}{\sqrt{r(1-r)}} I_{(0,1)}(r)$$

(Note:  $B_t^\alpha := \frac{1}{\sqrt{\alpha}} B_{\alpha t}$  is also a Brownian motion, and

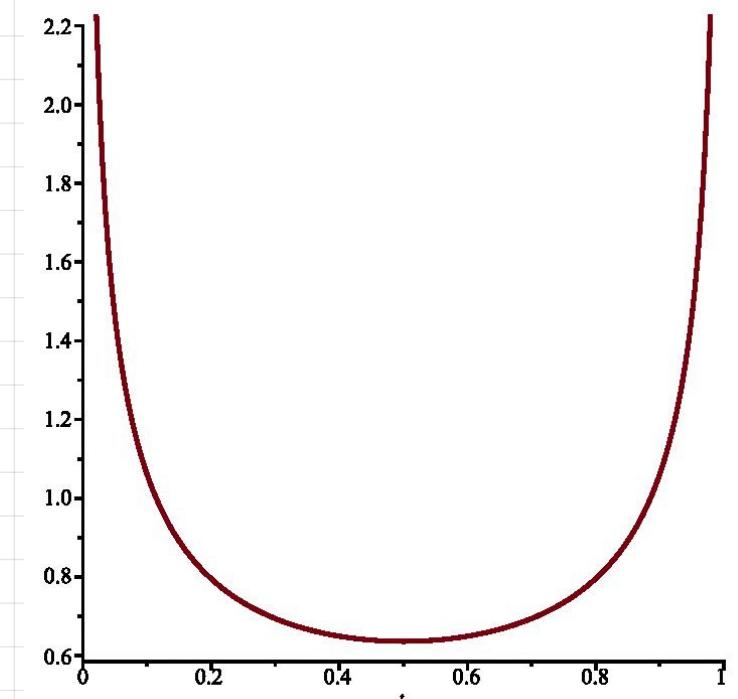
$$\begin{aligned} L_\alpha &\stackrel{d}{=} L_{\alpha}^{B^\alpha} = \sup\{t \leq \alpha : B_t^\alpha = 0\} = \sup\{t \leq \alpha : B_{\alpha t} = 0\} \\ &= \alpha \cdot \sup\{s \leq 1 : B_s = 0\} = \alpha L_1 \end{aligned}$$

$$\begin{aligned} \text{Pf. } P^*(L_1 \leq t) &= \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi t}} e^{-x^2/2t} P^*(T_0 > 1-t) dx \quad [\text{HW}] \\ &= P^*(T_x > 1-t) \end{aligned}$$

$$= 2 \cdot \frac{1}{\sqrt{2\pi t}} \int_0^\infty e^{-x^2/2t} dx \int_{1-t}^\infty \frac{x}{\sqrt{2\pi u^3}} e^{-x^2/2u} du$$

calculus

$$\begin{aligned} &= \frac{1}{\pi} \int_{1-t}^\infty \frac{\sqrt{t}}{\sqrt{u(t+u)}} du = \frac{1}{\pi} \int_0^t \frac{dr}{\sqrt{r(1-r)}}. \\ &\text{sub. } r = t/(u+1) \quad // \end{aligned}$$



More fun facts :

$$L_1 = \sup \{ 0 \leq t \leq 1 : B_t = 0 \} \stackrel{d}{=} \text{Arcsine Law}$$

$$d_{II} \\ M_1 = \inf \{ t > 0 : B_t = B_1^{\max} \}$$

$$\Lambda = |\{t \in [0, 1] : B_t > 0\}|$$

(See [Kallenberg, Thm. B.16].)