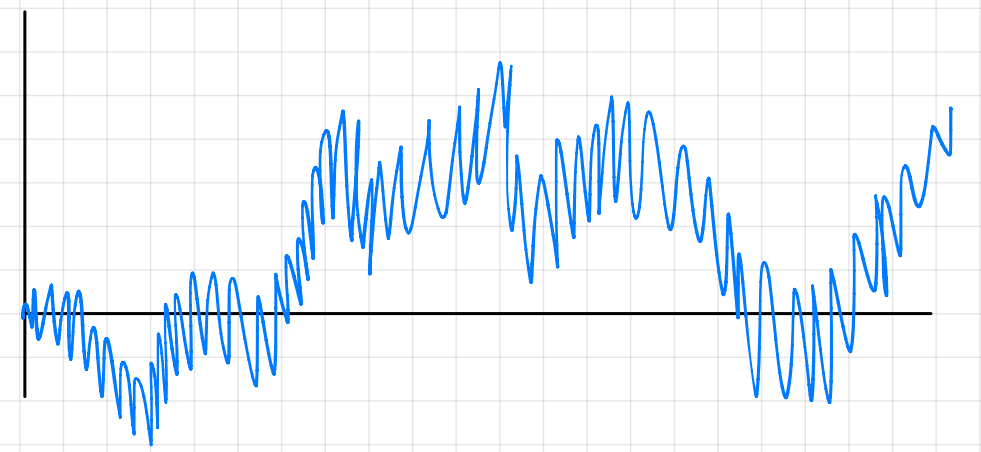


Let $(B_t)_{t \geq 0}$ be a Brownian motion on \mathbb{R} .

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

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) :

$$P(B_t^{\max} \geq z, B_t < z - y)$$

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

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

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

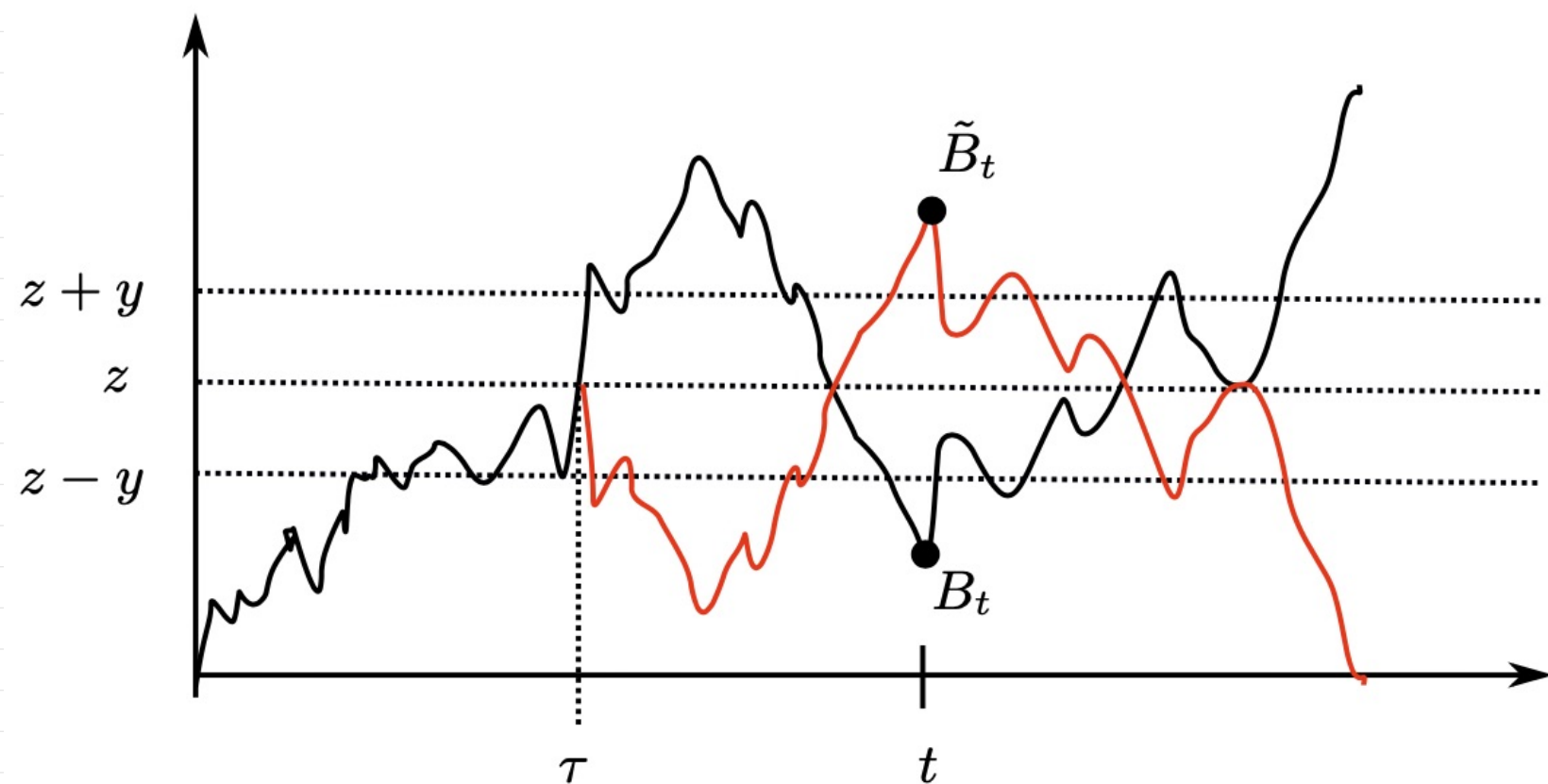
Let $(\tilde{B}_t)_{t \geq 0}$ be the Brownian motion reflected at time τ ,

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

Then on $\{\tau \leq t\}$, $\{B_t < z - y\}$

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

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



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

In particular,

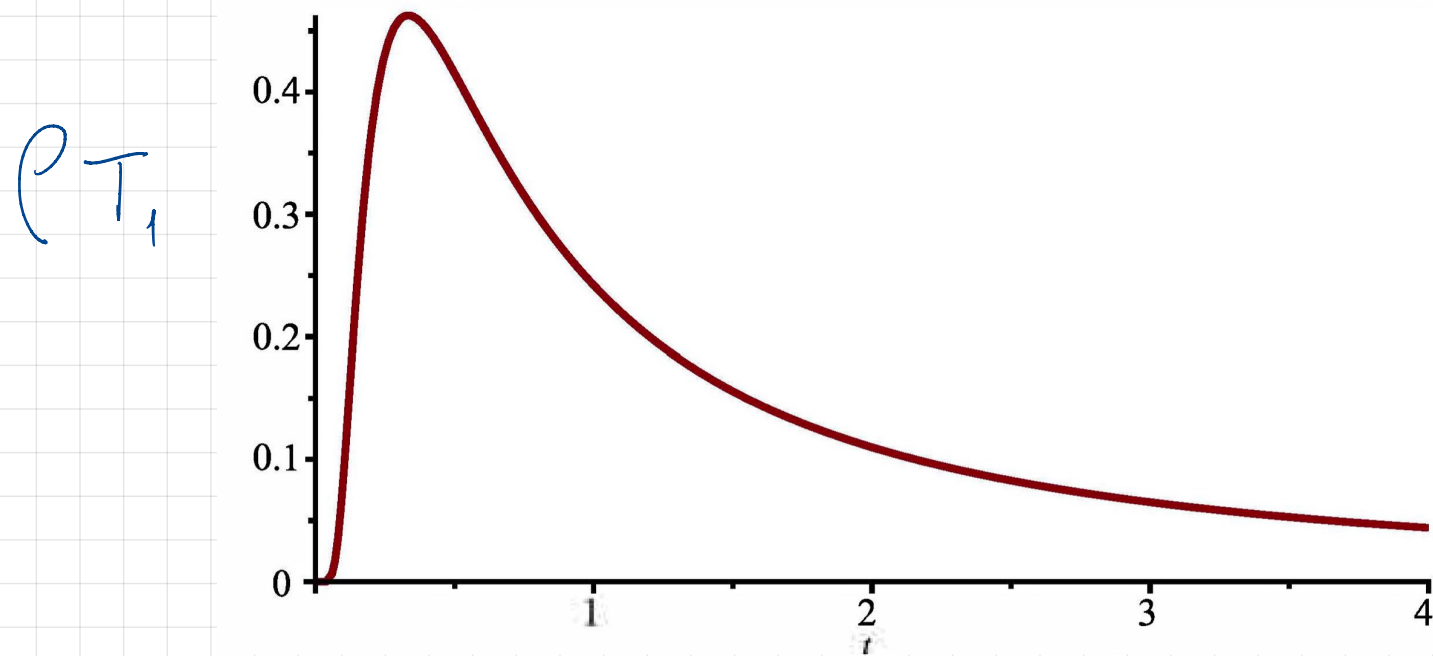
$$\mathbb{P}^0(B_t^{\max} \geq z) = \mathbb{P}^0(B_t^{\max} \geq z, B_t \geq z) + \mathbb{P}^0(B_t^{\max} \geq z, B_t < z)$$

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

$$\mathbb{P}^0(T_z \leq t)$$

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

$$p_{T_z}(t) = \frac{z}{\sqrt{2\pi t^3}} e^{-z^2/2t} \mathbb{1}_{(0, \infty)}(t).$$



In particular, $\mathbb{P}(T_z < \infty) = 1 \quad \forall z$

But

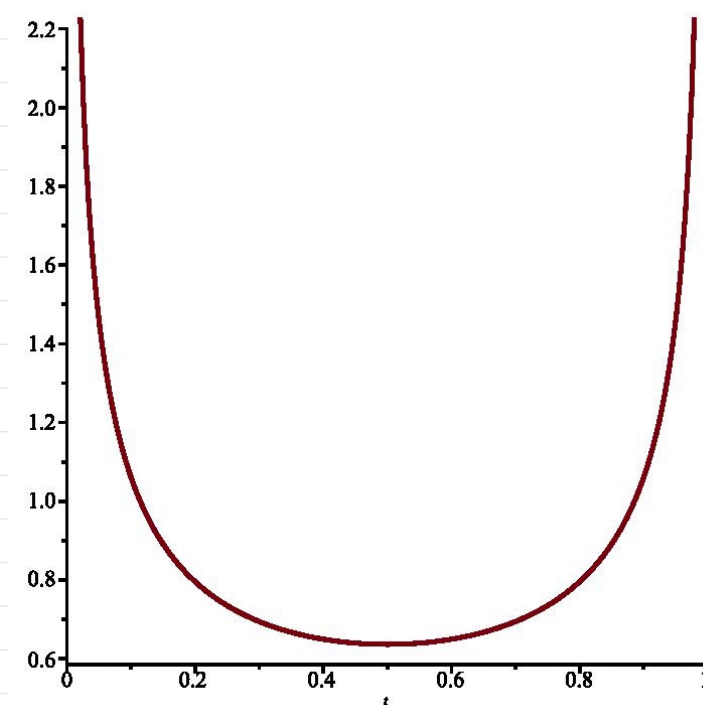
$$\begin{aligned} \mathbb{E}^0[T_z] &= \int_0^{\infty} t \cdot p_{T_z}(t) dt \\ &= \frac{z}{\sqrt{2\pi}} \int_0^{\infty} \frac{1}{\sqrt{t}} e^{-z^2/2t} dt \end{aligned}$$

We can use this density to compute other distributions exactly.

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

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

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



Pf. $P^\circ(L_1 \leq t) = \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi t}} e^{-x^2/2t} P^x(T_0 > 1-t) dx$ [HW]

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

More fun facts:

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

$d \parallel$

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

$d \parallel$

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

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