



Unbiased shifts of Brownian motion and balancing random measures

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joint work with

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1. Definition of an unbiased shift

Setting

$B = (B_t)_{t \in \mathbb{R}}$ is a two-sided standard Brownian motion starting in 0 ($B_0 = 0$) defined on its canonical probability space $(\Omega, \mathcal{A}, \mathbb{P}_0)$.

Definition

An **unbiased shift** (of B) is a random time T (negative values are allowed) such that:

- $B^{(T)} := (B_{T+t} - B_T)_{t \in \mathbb{R}}$ is a Brownian motion,
- $B^{(T)}$ is independent of B_T .

Example

If $T \geq 0$ is a stopping time, then $(B_{T+t} - B_T)_{t \geq 0}$ is a one-sided Brownian motion independent of B_T . However, the example

$$T := \inf\{t \geq 0: B_t = a\}$$

shows that $(B_{T+t} - B_T)_{t \in \mathbb{R}}$ need not be a two-sided Brownian motion.

Example

Consider a deterministic $T \equiv t_0$. Then $B^{(T)} = (B_{t_0+t} - B_{t_0})_{t \in \mathbb{R}}$ is a two-sided Brownian. However, since $B_{-t_0}^{(T)} = -B_{t_0}$, this two-sided motion is not independent of $B_T = B_{t_0}$.

Remark

An unbiased shift with $B_T = 0$ is characterized by

$$(B_{T+t})_{t \in \mathbb{R}} \stackrel{d}{=} B.$$

According to Mandelbrot ([The Fractal Geometry of Nature](#)) „...the process of Brownian zeros is stationary in a weakened form.“ He is using the (non-rigorous) concept of [conditional stationarity](#).

However, the stopping time

$$T := \inf\{t \geq 1 : B_t = 0\}$$

has the property $B_T = 0$. But clearly $B^{(T)}$ is not a Brownian motion. Something is missing!

Definition

Let ℓ^0 be the **local time** (random measure) at zero. Its right-continuous (generalised) inverse is defined as

$$T_r := \begin{cases} \sup\{t \geq 0 : \ell^0[0, t] = r\}, & r \geq 0, \\ \sup\{t < 0 : \ell^0[t, 0] = -r\}, & r < 0. \end{cases}$$

Theorem

Let $r \in \mathbb{R}$. Then T_r is an unbiased shift.

Idea of the proof: The intervals $[T_n, T_{n+1}]$, $n \in \mathbb{Z}$, split B into iid-cycles.

2. Local time

Definition

The local time measure ℓ^x at $x \in \mathbb{R}$ can be defined by

$$\ell^x(C) := \lim_{h \rightarrow 0} \frac{1}{h} \int \mathbf{1}\{s \in C, x \leq B_s \leq x + h\} ds.$$

Hence

$$\int f(B_s, s) ds = \iint f(x, s) \ell^x(ds) dx \quad \mathbb{P}_0\text{-a.s.}$$

for all measurable $f : \mathbb{R}^2 \rightarrow [0, \infty)$.

Definition

For $t \in \mathbb{R}$ the **shift** $\theta_t: \Omega \rightarrow \Omega$ is given by

$$(\theta_t \omega)_s := \omega_{t+s}, \quad s \in \mathbb{R}.$$

For $x \in \mathbb{R}$ let

$$\mathbb{P}_x := \mathbb{P}_0(B + x \in \cdot), \quad x \in \mathbb{R},$$

where B is the identity on Ω .

Remark

It is possible to choose a **perfect** version of local times, that is a (measurable) kernel satisfying for all $x \in \mathbb{R}$ and \mathbb{P}_x -a.e. that ℓ^x is diffuse and

$$\begin{aligned} \ell^x(\theta_t \omega, C - t) &= \ell^x(\omega, C), \quad C \in \mathcal{B}, t \in \mathbb{R}, \\ \ell^x(B, \cdot) &= \ell^0(B - x, \cdot). \end{aligned}$$

3. Palm measures and allocation rules

Definition

Let

$$\mathbb{P} := \int \mathbb{P}_x dx$$

be the distribution of a Brownian motion with a „uniformly distributed“ starting value.

Remark

A consequence of the stationary increments of B is the invariance property

$$\mathbb{P} = \mathbb{P} \circ \theta_s, \quad s \in \mathbb{R}.$$

Definition

A **random measure** ξ on \mathbb{R} is a kernel from Ω to \mathbb{R} such that $\xi(\omega, B) < \infty$ for \mathbb{P} -a.e. ω and all compact $B \subset \mathbb{R}$. It is called **invariant** if \mathbb{P} -a.e.:

$$\xi(\theta_t \omega, C - t) = \xi(\omega, C), \quad C \in \mathcal{B}, t \in \mathbb{R}.$$

In this case its **Palm measure** \mathbb{Q}_ξ is defined by

$$\mathbb{Q}_\xi(A) := \mathbb{E} \int \mathbf{1}_{[0,1]}(s) \mathbf{1}_A(\theta_s B) \xi(ds), \quad A \in \mathcal{A}.$$

If the **intensity** $\lambda_\xi := \mathbb{E}_\mathbb{P} \xi[0, 1] = \mathbb{Q}_\xi(\Omega)$ is positive and finite, then the Palm probability measure of ξ is defined by

$$\mathbb{Q}_\xi^0 := \lambda_\xi^{-1} \mathbb{Q}_\xi.$$

Theorem (Geman and and Horowitz '73)

The Palm (probability) measure of the local time ℓ^x is \mathbb{P}_x .

Definition

Let ν be a probability measure on \mathbb{R} . Define

$$\mathbb{P}_\nu := \int \mathbb{P}_x \nu(dx), \quad \ell^\nu := \int \ell^x \nu(dx).$$

Corollary

\mathbb{P}_ν is the Palm probability measure of ℓ^ν .

Remark

In the language of stochastic analysis ℓ^ν is a continuous **additive functional** with **Revuz measure** ν .

Definition

An **allocation rule** is a measurable mapping $\tau : \Omega \times \mathbb{R} \rightarrow \mathbb{R}$ that is **equivariant** in the sense that

$$\tau(\theta_t \omega, \mathbf{s} - t) = \tau(\omega, \mathbf{s}) - t, \quad \mathbf{s} \in \mathbb{R}, t \in \mathbb{R}, \mathbb{P}\text{-a.e. } \omega \in \Omega.$$

Theorem (L. and Thorisson '09)

Let ξ and η be two invariant random measures with positive and finite intensities. Let τ be an allocation rule and define $T := \tau(\cdot, 0)$. Then

$$\mathbb{Q}_\xi(\theta_T B \in \cdot) = \mathbb{Q}_\eta$$

iff τ is **balancing** ξ and η , that is

$$\int \mathbf{1}\{\tau(\mathbf{s}) \in \cdot\} \xi(d\mathbf{s}) = \eta \quad \mathbb{P}\text{-a.e.}$$

4. Unbiased shifts and balancing allocation rules

Definition (Skorokhod embedding problem)

Let ν be a probability measure on \mathbb{R} . A random time T **embeds** ν if B_T has distribution ν .

Theorem

Let T be a random time and ν be a probability measure on \mathbb{R} . Then T is an unbiased shift embedding ν if and only if the allocation rule τ defined by $\tau_T(s) := T \circ \theta_s + s$ is balancing ℓ^0 and ℓ^ν .

Example

Let $r > 0$. Then

$$\tau(s) := \inf\{t > s : \ell^0([s, t]) = r\}, \quad s \in \mathbb{R}.$$

Then τ is an allocation rule balancing ℓ^0 with itself. Hence $T_r = \tau(\cdot, 0)$ is an unbiased shift (embedding δ_0).

5. Existence of unbiased shifts

Theorem

Let ν be a probability measure on \mathbb{R} with $\nu\{0\} = 0$. Then the stopping time

$$T := \inf\{t > 0: \ell^0[0, t] = \ell^\nu[0, t]\}$$

embeds ν and is an unbiased shift.

Remark

The above stopping time above was introduced in Bertoin and Le Jan (1992) as a solution of the Skorokhod embedding problem.

Theorem

Let ν be a probability measure on \mathbb{R} . Then there exists a nonnegative stopping time that is an unbiased shift embedding ν .

Theorem

Assume that $\nu\{0\} < 1$. Then any unbiased shift T embedding ν satisfies

$$\mathbb{P}_0\{T = 0\} = 0.$$

If, however, $\nu = \delta_0$, then for any $p \in [0, 1]$ there is an unbiased shift $T \geq 0$ embedding ν and such that $\mathbb{P}_0\{T = 0\} = p$.

Theorem

Let ξ and η be jointly stationary orthogonal diffuse random measures on \mathbb{R} with finite and equal intensities. Then the mapping $\tau: \Omega \times \mathbb{R} \rightarrow \mathbb{R}$, defined by

$$\tau(s) := \inf\{t > s: \xi[s, t] = \eta[s, t]\}, \quad s \in \mathbb{R},$$

is an allocation rule balancing ξ and η .

Remark

The previous theorem holds in a more general stationary setting. The assumption of equal intensities has to be replaced by

$$\mathbb{E}[\xi[0, 1]|\mathcal{I}] = \mathbb{E}[\eta[0, 1]|\mathcal{I}] \quad \mathbb{P}\text{-a.e.},$$

where \mathcal{I} is the **invariant σ -field**. In the Brownian setting, \mathbb{P} is trivial on \mathcal{I} . (If $A \in \mathcal{I}$ then either $\mathbb{P}(A) = 0$ or $\mathbb{P}(A^c) = 0$.)

6. Moment properties of unbiased shifts

Theorem

If T is an unbiased shift embedding a probability measure $\nu \neq \delta_0$, then

$$\mathbb{E}_0 \sqrt{|T|} = \infty.$$

Idea of the proof:

- Take an $x > 0$ such that $\nu[x, \infty) = \mathbb{P}(B_T > x) > 0$.
- On the event $\{B_T > x\}$, T can be bounded from below by the minimum of two independent hitting times for $-x$, independent of B_T .
- Use the moment properties of hitting times.

Theorem

Suppose ν is a distribution with $\nu\{0\} = 0$, and finite absolute mean. If the stopping time $T \geq 0$ is an unbiased shift embedding ν , then

$$\mathbb{E}_0 T^{1/4} = \infty.$$

Theorem

Let ν be as above and let T be the Bertoin/Le Jan stopping time. Then, for all $\beta \in [0, 1/4)$,

$$\mathbb{E}_0 T^\beta < \infty.$$

Theorem

Let $T \geq 0$ be an unbiased shift embedding δ_0 . If T is non-trivial, that is $\mathbb{P}_0\{T = 0\} < 1$, then

$$\mathbb{E}_0 T = \infty.$$

Example

There is an unbiased shift $T \neq 0$ embedding δ_0 , such that $\mathbb{E}e^{\lambda|T|} < \infty$ for some $\lambda > 0$.

7. Minimality

Definition

A random time $T \geq 0$ is called **minimal unbiased shift** if it is an unbiased shift and if for any other unbiased shift S embedding $\mathbb{P}(B_T \in \cdot)$ and such such that $\mathbb{P}_0(0 \leq S \leq T) = 1$ we have that $\mathbb{P}_0(S = T) = 1$.

Theorem

The Bertoin-Le Jan stopping time T is a minimal unbiased shift.

8. References

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