



Invariance principle for random walks with anomalous recurrence properties

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Joint work with
Nicholas Georgiou and Andrew Wade

Outline

- 1 From classical to nonhomogeneous random walk
- 2 Elliptical random walk
- 3 Diffusion limits

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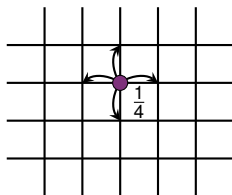
Classical zero-drift random walks

1. Symmetric simple random walk on \mathbb{Z}^d

- $X_n \in \mathbb{Z}^d$, $X_0 = 0$.
- Given X_0, \dots, X_n , new location X_{n+1} is uniformly distributed on the $2d$ adjacent lattice sites to X_n .

Theorem (Pólya 1921)

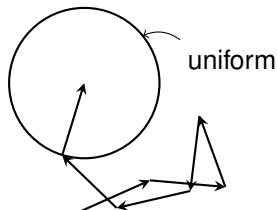
SRW is recurrent if $d = 1$ or $d = 2$, but transient if $d \geq 3$.



Classical zero-drift random walks

2. Pearson–Rayleigh random walk in \mathbb{R}^d

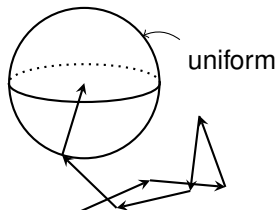
- $X_n \in \mathbb{R}^d$, $X_0 = 0$.
- Given X_0, \dots, X_n , new location X_{n+1} is uniformly distributed on the unit circle/sphere centred at X_n .



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Recurrence/transience of homogeneous random walks

Let (X_n) be a **spatially homogeneous** random walk in \mathbb{R}^d .

So X_{n+1} depends only on X_n , but $\Delta := X_{n+1} - X_n$ is independent of X_n (and n).

Let $\mu = \mathbb{E}\Delta$, the **mean drift** vector of the random walk.

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Theorem (Chung–Fuchs)

Under mild conditions, if $\mu = 0 \in \mathbb{R}^d$, then (X_n) is

- *recurrent if $d = 1$ or $d = 2$;*
- *transient if $d \geq 3$.*

This result applies both to the symmetric simple RW and the Pearson–Rayleigh RW.

Definition

- recurrence: $\mathbb{P}[\text{return to (nbrhood of) origin}] = 1$.
- transience: $\mathbb{P}[\text{return to (nbrhood of) origin}] < 1$.

Scaling limit for homogeneous random walks

Under mild non-degeneracy conditions (non-singularity of $\mathbb{E}[\Delta\Delta^\top]$), we have (up to a linear transformation):

Theorem (Donsker)

*Spatially homogeneous random walk in \mathbb{R}^d with zero drift converges to d -dimensional **Brownian motion** after diffusive scaling:*

$$\left(\frac{X_{\lfloor nt \rfloor}}{\sqrt{n}} \right)_{t \in [0,1]} \implies (b_t)_{t \in [0,1]}.$$

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Brownian motion on \mathbb{R}^d ($d \geq 2$) possesses a **skew-product representation**.

Let $r_t := \|b_t\|$, $\theta_t := \frac{b_t}{\|b_t\|}$. Then,

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Non-homogeneous random walks

What if we allow $\Delta = X_{n+1} - X_n$, the **jump distribution**, to depend on the current location?

Then $\mu(x) := \mathbb{E}_x \Delta := \mathbb{E}[\Delta \mid X_n = x]$ becomes a function of the current position $x \in \mathbb{R}^d$.

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Question

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Answer

For $d = 1$: **yes** (essentially) — zero drift implies recurrence.
For higher dimensions: **no** — either behaviour is possible.

Theorem

There exist non-homogeneous random walks with $\mu(x) = 0$ for all $x \in \mathbb{R}^d$ that are

- *transient in $d = 2$;*
- *recurrent in $d \geq 3$.*

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Elliptical random walk (in \mathbb{R}^2)

We modify the Pearson–Rayleigh random walk to make jumps distributed on an **ellipse**.

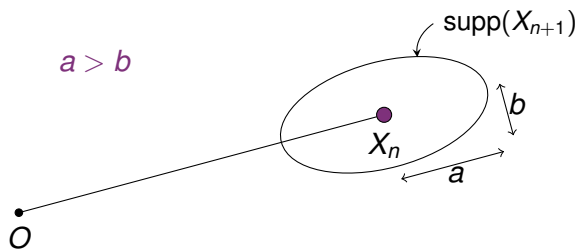
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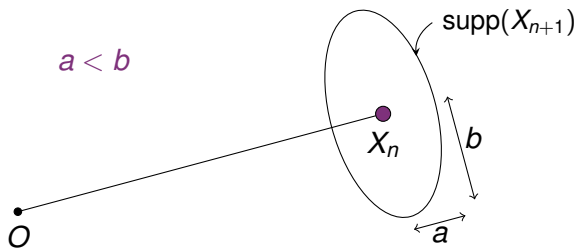


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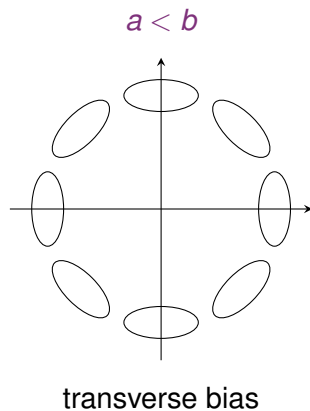
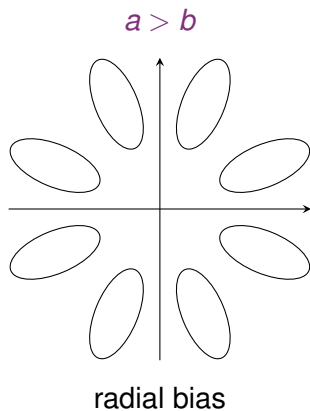
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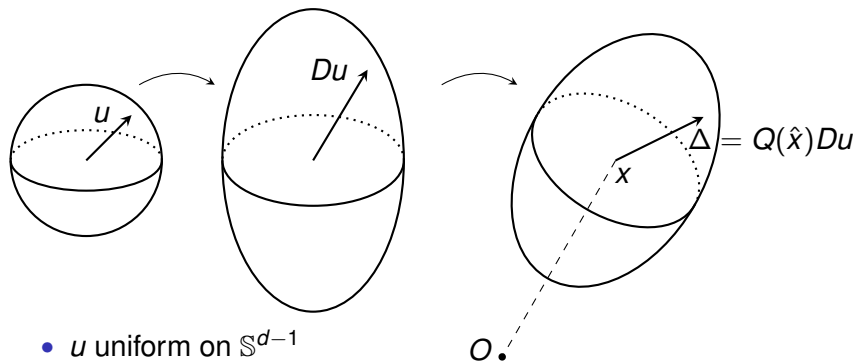


Elliptical random walk



Elliptical random walk ($d \geq 2$)

Suppose $X_n = x \in \mathbb{R}^d$. Write \hat{x} for unit vector in direction x .



- u uniform on \mathbb{S}^{d-1}
- $D = \text{diag}(a, b, \dots, b)$
- $Q(\hat{x})$ orthogonal matrix, with $Q(\hat{x})e_1 = \hat{x}$.

Moments of Δ

Notation: write $\mathbb{E}_x[\cdot]$ for $\mathbb{E}[\cdot \mid X_n = x]$ and write Δ_x for the component of Δ in direction x :

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Therefore, by construction,

$$\mathbb{E}_x[\Delta] = 0, \quad \mathbb{E}_x[\Delta\Delta^\top] = \frac{1}{d}Q(\hat{x})D^2Q^\top(\hat{x}).$$

Hence,

$$\mathbb{E}_x[\Delta_x] = 0, \quad \mathbb{E}_x[\Delta_x^2] = \frac{a^2}{d}, \quad \mathbb{E}_x[\|\Delta\|^2] = \frac{a^2 + (d-1)b^2}{d}.$$

Radial component of X_n

We analyse (X_n) by considering $R_n := \|X_n\|$.

By symmetry, R_n is also Markov (R_n is a non-homogeneous random walk on \mathbb{R}_+).

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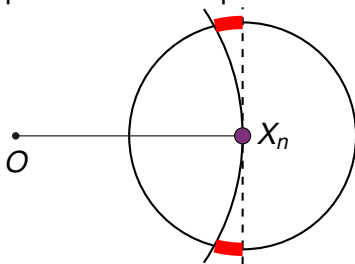
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Moreover, it has asymptotically zero drift:

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where positive constant c depends on model parameters and ambient **dimension**.



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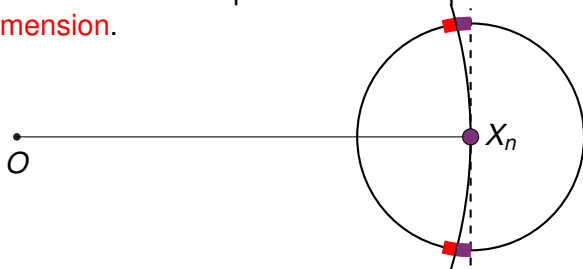
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Lamperti's classification

Define $\mu_k(r) := \mathbb{E}[(R_{n+1} - R_n)^k \mid R_n = r]$.

In the early 1960s, John Lamperti studied in detail how the asymptotics of a stochastic process on \mathbb{R}_+ are determined by the first two moment functions of its increments, μ_1 and μ_2 .

Theorem (Lamperti, 1960)

Let (R_n) be a Markov chain on \mathbb{R}_+ . Under mild conditions:

- If $2r\mu_1(r) - \mu_2(r) > 0$ for all large enough r , then R_n is *transient*,
- If $2r\mu_1(r) - \mu_2(r) < 0$ for all large enough r , then R_n is *recurrent*.

Recurrence/transience of elliptical random walk

Given $X_n = x$,

$$\begin{aligned}R_{n+1} - R_n &= \|x + \Delta\| - \|x\| \\&= [\dots \text{expand using Taylor's theorem} \dots] \\&= \Delta_x + \frac{\|\Delta\|^2 - \Delta_x^2}{2\|x\|} + O(\|x\|^{-2}).\end{aligned}$$

So,

$$\mu_1(r) = \frac{(d-1)b^2}{d} \frac{1}{2r} + O(r^{-2}), \quad \mu_2(r) = \frac{a^2}{d} + O(r^{-1}).$$

Theorem

Let (X_n) be an elliptical random walk in \mathbb{R}^d , with parameters a and b .

- If $(d-1)b^2 - a^2 > 0$ then (X_n) is **transient**.
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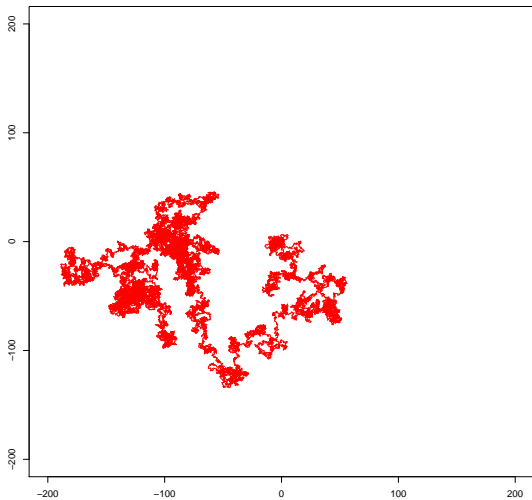
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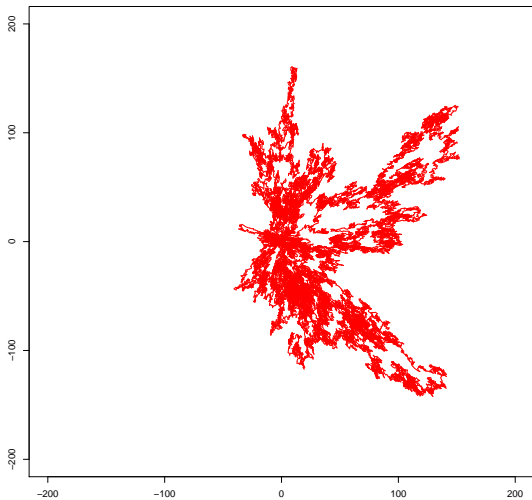
Simulations

$a = 1, b = 1$



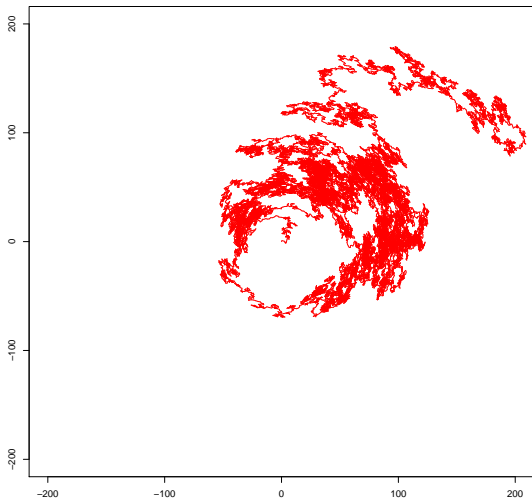
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$a = 2, b = 1$



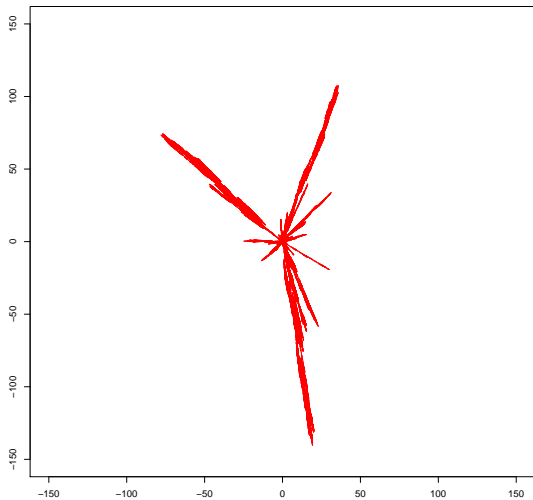
Simulations

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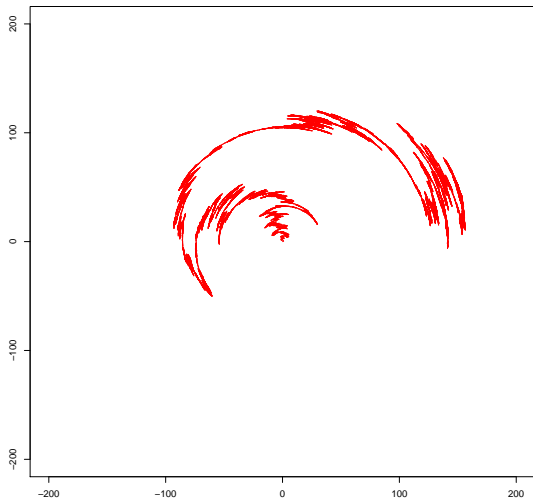
Simulations

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Simulations

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Diffusion limits

Back to homogeneous case:

Theorem (Donsker)

The Pearson–Rayleigh walk in \mathbb{R}^d (the case $a \equiv b = 1$) converges to d -dimensional Brownian motion:

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Now, more generally:

Theorem

If (X_n) is an elliptical random walk in \mathbb{R}^d , then there exists a continuous strong Markov process (a diffusion) (\mathcal{X}_t) on \mathbb{R}^d , whose law depends on the parameters a and b , such that,

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Brownian motion and Bessel processes

Brownian motion on \mathbb{R}^d ($d \geq 2$) possesses a **skew-product representation**.

Let $r_t := \|b_t\|$, $\theta_t := \frac{b_t}{\|b_t\|}$. Then,

- r_t is a **Bessel process** on \mathbb{R}_+ of ‘dimension’ (parameter) d ;
- θ_t is a (stochastic) time-change of an **independent** Brownian motion on the **sphere**.

- A Bessel process with ‘dimension’ δ , $\text{BES}(\delta)$, is a Markov process β_t on \mathbb{R}_+ satisfying the SDE

$$d\beta_t = \frac{\delta - 1}{2\beta_t} \mathbf{1}_{\{\beta_t \neq 0\}} dt + dW_t,$$

where W_t is BM on \mathbb{R} .

- $0 \in \mathbb{R}_+$ is recurrent for $\text{BES}(\delta)$ if $1 \leq \delta < 2$ and transient if $\delta \geq 2$.

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- Define the additive functional $\rho(t) := \int_0^t r_s^{-2} ds$.
- Then $\theta_t = \varphi_{\rho(t)}$, where φ_t is BM on \mathbb{S}^{d-1} **independent** of r_t .
- That is, φ_t solves the SDE

$$d\varphi_t = -\frac{d-1}{2}\varphi_t dt + (I - \varphi_t \varphi_t^\top) dW_t,$$

where W_t is BM on \mathbb{R}^d .

Diffusion limit of elliptical random walk

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- θ_t is a time-change of a two-sided BM $(\varphi_t)_{t \in \mathbb{R}}$ on \mathbb{S}^{d-1} , **independent** of r_t .

General setting for invariance principle

- **Moments condition:** $\sup_x \mathbb{E}_x[\|\Delta\|^4] < \infty$.
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- **Asymptotic isotropy:** $M(x) \rightarrow \sigma^2(\hat{x})$ as $\|x\| \rightarrow \infty$ for a positive-definite matrix valued C^∞ -function σ^2 on \mathbb{S}^{d-1} .

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Define for each $u \in \mathbb{S}^{d-1}$ an inner product $\langle \cdot, \cdot \rangle_u$ on \mathbb{R}^d via

$$\langle y, z \rangle_u := y^\top \cdot \sigma^2(u) \cdot z = \langle y, \sigma^2(u) \cdot z \rangle, \quad (\text{for } y, z \in \mathbb{R}^d).$$

- **Limiting covariance regularity:** There exist constants $U, V, \delta > 0$ such that, for all $u, v \in \mathbb{S}^{d-1}$,

$$\langle u, u \rangle_u = U, \quad \text{tr } \sigma^2(u) = V, \quad \text{and } \langle v, v \rangle_u \geq \delta.$$

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- **Limiting radial structure:** $u \in \mathbb{S}^{d-1}$ is eigenvector of $\sigma^2(u)$.

General setting for invariance principle

Theorem

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$$\left(\frac{X_{\lfloor nt \rfloor}}{\sqrt{n}} \right)_{t \in [0,1]} \Longrightarrow (\mathcal{X}_t)_{t \in [0,1]}.$$

The diffusion (\mathcal{X}_t) is the unique weak solution of the SDE

$$d\mathcal{X}_t = \sigma(\hat{\mathcal{X}}_t) dW_t, \quad \mathcal{X}_0 = 0,$$

where W is BM on \mathbb{R}^d and σ any square-root of σ^2 .

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$$d\mathcal{X}_t = \sigma(\hat{\mathcal{X}}_t) dW_t, \quad \mathcal{X}_0 = 0,$$

where W is BM on \mathbb{R}^d and σ any square-root of σ^2 .

Typically $x \mapsto \sigma(\hat{x})$ has a discontinuity at $0 \in \mathbb{R}^d$ and (\mathcal{X}_t) keeps visiting 0, so standard methods from (Ethier & Kurtz, 1986) need to be extended (key fact: Bessel local time at 0 vanishes).

General setting for invariance principle

Theorem

The martingale problem

$$d\mathcal{X}_t = \sigma(\hat{\mathcal{X}}_t)dW_t, \quad \text{for any deterministic } \mathcal{X}_0 \in \mathbb{R}^d,$$

is well-posed for any square-root σ of the asymptotic covariance structure σ^2 .

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- Excursion theory for (\mathcal{X}_t) has to be developed for *uniqueness in law* (works for any square-root σ).
- *Strong existence* and *pathwise uniqueness* may fail even for smooth σ (depends on the choice of square-root).

General setting: the excursion skew-decomposition

Let $(\xi_t, t \geq 0)$ be BM on \mathbb{R}^d . Then SDE

$$d\psi_t = (\sigma(\psi_t) - \psi_t \psi_t^\top) d\xi_t - \frac{V-1}{2} \psi_t dt, \quad \psi_0 \in \mathbb{S}^{d-1}, \quad (1)$$

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Theorem

- (a) *Radial component.* The process r , defined by $r_t = \|\mathcal{X}_t\|$, is BES(U/V) started at 0.
- (b) *Skew-product structure.* Let $s > 0$ and $\tau_s := \inf\{t \geq s : r_t = 0\}$. Then for any $t \in [s, \tau_s)$,

$$\hat{\mathcal{X}}_t = \varphi_{\rho_s(t)}, \quad \text{where } \rho_s(t) = \int_s^t r_u^{-2} du,$$

processes φ and r are independent and φ follows SDE (1) started according its stationary measure μ .

Properties of \mathcal{X}

Scaling: \mathcal{X} and $\mathcal{Y} = (c^{-1/2}\mathcal{X}_{ct})$, $c > 0$, have the same law:
$$d\mathcal{Y}_t = c^{-1/2}d\mathcal{X}_{ct} = c^{-1/2}\sigma(\hat{\mathcal{X}}_{ct})dW_{ct} = \sigma(\hat{\mathcal{Y}}_t)d(c^{-1/2}W_{ct})$$

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Rapid spinning: Let $s > 0$ and $\tau_s^- = \sup\{t < s : r_t = 0\}$. For any $t \in (\tau_s^-, \tau_s)$ in excursion interval, it holds

$$\lim_{s \downarrow \tau_s^-} \rho_s(t) = \infty, \quad \text{where} \quad \rho_s(t) = \int_s^t r_u^{-2} du. \quad (2)$$

Rapid spinning implies that $\hat{\mathcal{X}}_t = \varphi_{\rho_s(t)}$ is distributed according to the stationary measure μ of SDE (1).

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Proof of (2): (Pitman & Yor, 1982) BES(U/V) excursion (recall $\delta = U/V \in (1, 2)$): pick maximum according to σ -finite measure $m^{3-\delta} dm$ and run back-to-back two independent BES($4 - \delta$) from 0 it hits m . Apply (M & Urusov, 2012).

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where Δ_g is the Laplace–Beltrami operator on (\mathbb{S}^{d-1}, g) and vector field b is **explicit** in σ^2 and the metric g . Here,

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$$b = \frac{1}{2} \left((d - V)x_i - \frac{\partial \sigma_{ij}^2}{\partial x_j} + \frac{1}{2} g^{ik} g_{jl} \frac{\partial \sigma_{jl}^2}{\partial x_k} \right) \frac{\partial}{\partial x_i}.$$

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$\exists!$ invariant measure μ on \mathbb{S}^{d-1} , such that $\mu(dx) = \nu(x)dx$.

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Hence excursion representation for BES(U/V) (in \mathbb{R}_+) from (Pitman & Yor, 1982) extends to \mathcal{X} (in \mathbb{R}^d).

Some remarks

Walsh's Brownian motions: degenerate case $U = V$ is excluded from our results. But for U very close to V the measure on \mathbb{S}^d from Walsh's construction is our stationary measure angular measure μ of ψ . Heuristically this approximates Walsh's Brownian motion (recall simulation).

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Pathwise uniqueness and strong solutions of

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Since the solution is unique in law, the dichotomy is

- (i) pathwise uniqueness holds (implying strong uniqueness);
- (ii) pathwise uniqueness fails and the SDE has multiple solutions, none of which are strong.

Which of (i) or (ii) occurs **does** depend on the choice of square-root σ (e.g. multidimensional Tanaka SDE).

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\exists **smooth** σ under (ii) (including “complex Brownian motion” (Stroock & Yor, 1981). We have examples for $d = 2, 4, 8$.

Marginal limit theorem

At time $t = 1$, the law of \mathcal{X}_1 is given by

- $\|\mathcal{X}_1\|^2 \sim \Gamma(\frac{1}{2} + (d-1)\frac{b^2}{2a^2}, 2a^2)$ (Gamma);
- $\hat{\mathcal{X}}_1 \sim U(\mathbb{S}^{d-1})$ (uniform);
- $\|\mathcal{X}_1\|$ and $\hat{\mathcal{X}}_1$ are independent.

(When $a = b$ then $\|\mathcal{X}_1\|^2$ is a scalar multiple of a χ^2 random variable with d degrees of freedom.)

So for example we get an angular ergodic result for the **random walk**: for measurable $A \subseteq \mathbb{S}^{d-1}$,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} \mathbf{1}\{\hat{X}_k \in A\} = \frac{|A|}{|\mathbb{S}^{d-1}|}, \text{ in } L^1.$$

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Almost-sure version unlikely to hold as the limit is non-degenerate

$$\int_0^1 \mathbf{1}\{\hat{\mathcal{X}}_t \in \mathfrak{A}\} dt = \lim_{\varepsilon \downarrow 0} \int_\varepsilon^1 \mathbf{1}\{\hat{\mathcal{X}}_t \in \mathfrak{A}\} dt.$$

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