

Bessel Random Walks for Identities of Higher-order Bernoulli and Euler Polynomials

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Euler polynomials

Generating function

$$\sum_{n=0}^{\infty} E_n(x) \frac{z^n}{n!} = \frac{2}{e^z + 1} e^{xz}$$

and

$$\sum_{n=0}^{\infty} E_n^{(p)}(x) \frac{z^n}{n!} = \left(\frac{2}{e^z + 1} \right)^p e^{xz}$$

$$E_n^{(p)}(x) = \sum_{k_1 + \dots + k_p + k = n} \binom{n}{k_1, \dots, k_p, k} x^k E_{k_1}(0) E_{k_2}(0) \cdots E_{k_p}(0).$$

$$E_n(x) = P \left(E_{n_1}^{(p_1)}(x), \dots, E_{n_k}^{(p_k)}(x) \right)?$$

Theorem(L. Jiu, V. H. Moll and C. Vignat)

For any positive integer N ,

$$E_n(x) = \frac{1}{N^n} \sum_{\ell=N}^{\infty} p_{\ell}^{(N)} E_n^{(\ell)} \left(\frac{\ell - N}{2} + Nx \right),$$

where

$$\frac{1}{T_N(1/z)} = \sum_{\ell=0}^{\infty} p_{\ell}^{(N)} z^{\ell},$$

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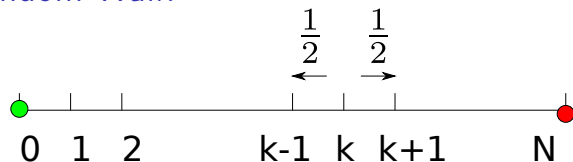
where

$$\frac{1}{T_N(1/z)} = \sum_{\ell=0}^{\infty} p_{\ell}^{(N)} z^{\ell}, \quad T_N(\cos \theta) = \cos(N\theta)$$

$N = 2$: $T_2(z) = 2z^2 - 1$ and $\frac{1}{T_2(1/z)} = \frac{z^2}{2-z^2}$

$$E_n(x) = \frac{1}{2^n} \sum_{\ell=2}^{\infty} p_{\ell}^{(2)} E_n^{(\ell)} \left(\frac{\ell}{2} - 1 + 2x \right), \quad \text{where } p_{\ell}^{(2)} = \begin{cases} \frac{1}{2^k}, & \text{if } \ell = 2k; \\ 0, & \text{otherwise} \end{cases}$$

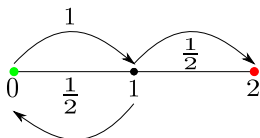
Random Walk



- ▶ 0 is the **source** and N is the **sink**;
- ▶ at each $k = 1, \dots, N - 1$, it is a “fair coin” walk;
- ▶ let ν_N be the random number of steps for this process.

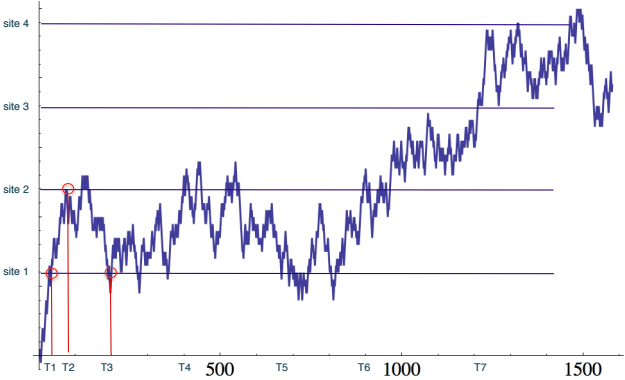
$$p_\ell^{(N)} = \mathbb{P}(\nu_N = \ell)$$

$N = 2$:

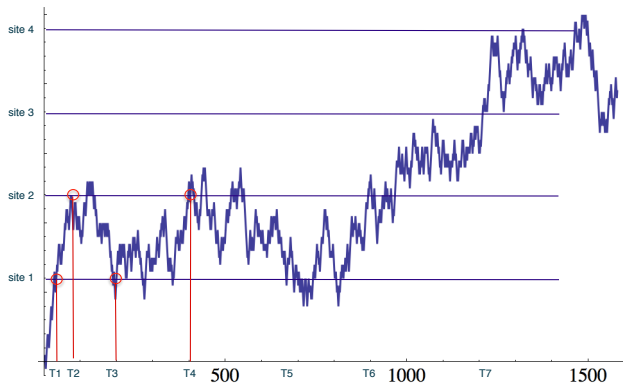


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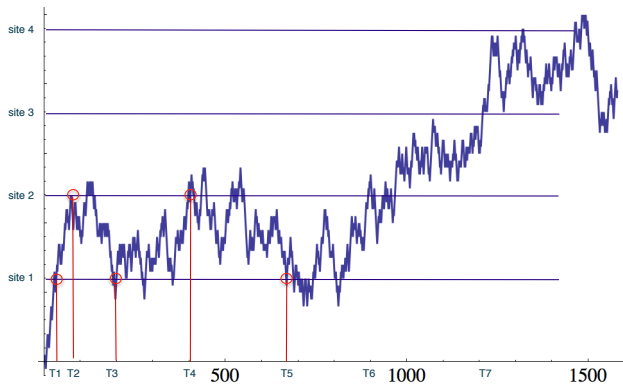
Reflected Brownian Motion



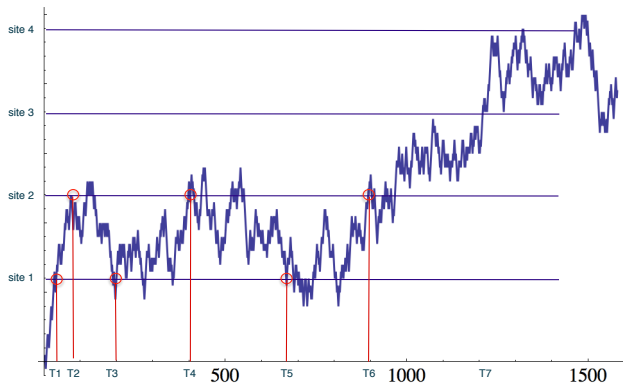
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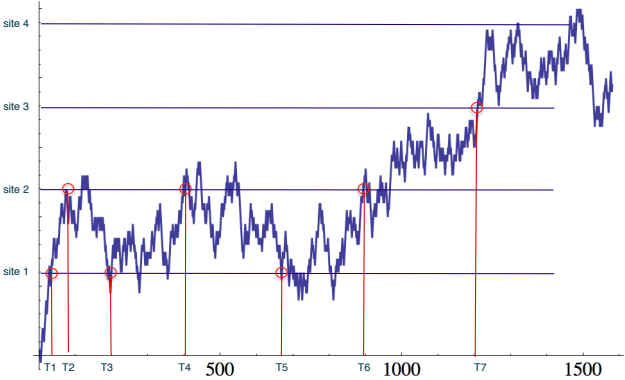
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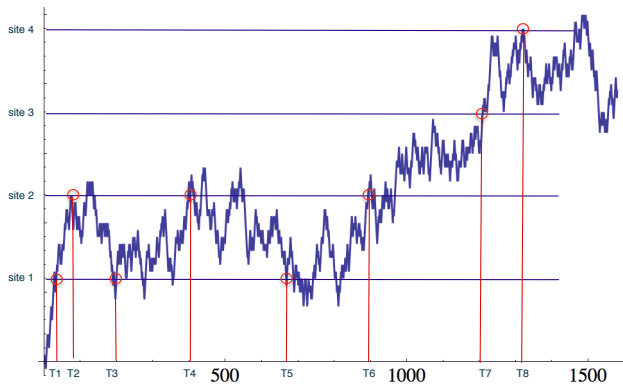
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Reflected Brownian Motion



Random Sum

- ▶ L_j are independent and identically distributed with **hyperbolic secant density**

$$\mathbb{E} [L_j^n] = \int_{\mathbb{R}} t^n \operatorname{sech}(\pi t) dt;$$

- ▶ ν_N is an integer valued random variable **independent** of the L_j 's:

$$\mathbb{E} [z^{\nu_N}] = \frac{1}{T_N\left(\frac{1}{z}\right)},$$

Theorem. [Klebanov et al.]. The random variable

$$Z_N = \frac{1}{N} \sum_{j=1}^{\nu_N} L_j$$

has **the same hyperbolic secant distribution** (as L_j 's).

Hyperbolic Secant

Let $L \sim \text{sech}(\pi t)$, then the Euler polynomial is given by

$$E_n(x) = \mathbb{E} \left[\left(x + iL - \frac{1}{2} \right)^n \right] = \int_{\mathbb{R}} \left(x + it - \frac{1}{2} \right)^n \text{sech}(\pi t) dt.$$

The proof uses the integral:

$$\int_{\mathbb{R}} t^k \text{sech}(\pi t) dt = \frac{|E_k|}{2^k}. \quad \left(\mathbb{E} \left[z^{x+iL-\frac{1}{2}} \right] = \frac{2}{e^z + 1} e^{xz} \right)$$

Let $\{L_j\}_{1 \leq j \leq p}$ be p independent random variables $L_j \sim \text{sech}(\pi t)$.
The generalized Euler polynomial is given by

$$E_n^{(p)}(x) = \mathbb{E} \left[\left(x + \left(iL_1 - \frac{1}{2} \right) + \cdots + \left(iL_p - \frac{1}{2} \right) \right)^n \right].$$

L. B. Klebanov, A. V. Kakosyan, S. T. Rachev, and G. Temnov. On a class of distributions stable under random summation. J. Appl. Prob., 49:303–318, 2012.

Probabilistic Interpretation

$$E_n(x) = \frac{1}{N^n} \sum_{\ell=N}^{\infty} p_{\ell}^{(N)} E_n^{(\ell)} \left(\frac{\ell - N}{2} + Nx \right),$$

$$\begin{aligned} L \sim \frac{1}{N} \sum_{j=1}^{\nu_N} L_j &\Rightarrow x + iL - \frac{1}{2} \sim x + \left(\frac{1}{N} \sum_{j=1}^{\nu_N} iL_j \right) - \frac{1}{2} \\ &\Rightarrow x + iL - \frac{1}{2} \sim \frac{1}{N} \sum_{j=1}^{\nu_N} \left(iL_j - \frac{\nu_N}{2} + Nx - \frac{N}{2} + \frac{\nu_N}{2} \right) \end{aligned}$$

Take moments.

Hitting Time

Consider

- ▶ a linear Brownian motion W_t starting from 0
- ▶ the hitting time T by W_t of level $z = 1$
- ▶ another independent Brownian motion ω_t .

Then

$$\omega_T \sim \text{sech}(x).$$

Denote

$$T_1 < T_2 < \dots < T_l = T$$

the successive epochs at which W_t visits the sites $z_i = \frac{i}{N}$, $0 \leq i \leq N$.

Hitting Time

This defines a random walk with

$$p_\ell^{(N)} = \mathbb{P}\{W_t \text{ reach the sink in } \ell \text{ steps}\}.$$

Now write

$$T = (T - T_{\ell-1}) + (T_{\ell-1} - T_{\ell-2}) + \cdots + (T_1 - 0)$$

and

$$\omega_T \sim \omega_{T-T_{\ell-1}} + \omega_{T_{\ell-1}-T_{\ell-2}} + \cdots + \omega_{T_1-0},$$

each term $\sim \text{sech}(x)$.

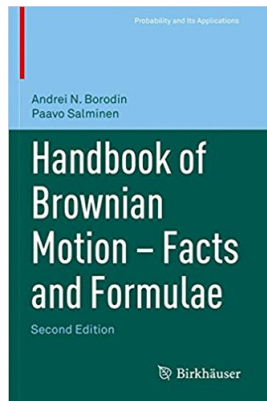
This corresponds Klebanov's [random sum decomposition](#)

$$Z_N = \frac{1}{N} \sum_{j=1}^{\nu_N} L_j.$$

Generalization

- ▶ Bessel process in \mathbb{R}^n :

$$R_t^{(n)} := \sqrt{(W_t^{(1)})^2 + \dots + (W_t^{(n)})^2}$$



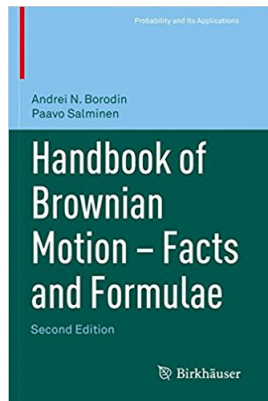
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$$\begin{aligned} & \mathbb{E}_x \left(e^{-\alpha H_z}; \sup_{0 \leq s \leq H_z} R_s^{(n)} < y \right) \\ &= \begin{cases} \frac{x^{-\nu} I_\nu(x\sqrt{2\alpha})}{z^{-\nu} I_\nu(z\sqrt{2\alpha})}, & 0 \leq x \leq z \leq y; \\ \frac{S_\nu(y\sqrt{2\alpha}, x\sqrt{2\alpha})}{S_\nu(y\sqrt{2\alpha}, z\sqrt{2\alpha})}, & z \leq x \leq y, \end{cases} \quad (2.1.4) \end{aligned}$$



Generalization

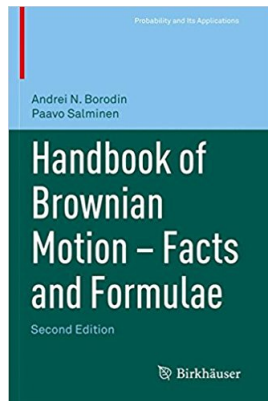
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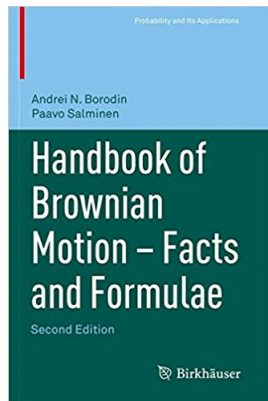
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- ▶ $n = 2 + 2\nu$ for $\nu \geq 0$

$$S_\nu(x, y) := (xy)^{-\nu} [I_\nu(x)K_\nu(y) - K_\nu(x)I_\nu(y)].$$



$$n = 3 \Leftrightarrow \nu = 1/2$$



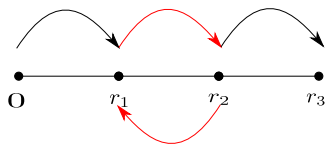
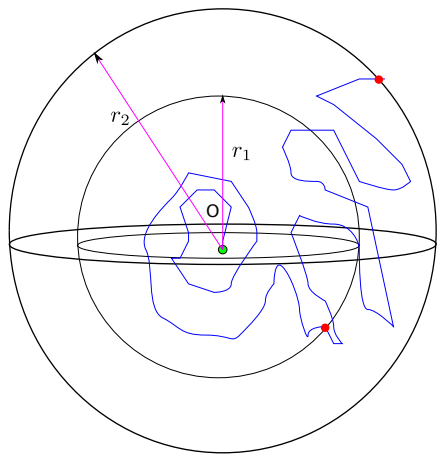
$$I_{\frac{1}{2}}(x) = \sum_{m=0}^{\infty} \frac{\left(\frac{x}{2}\right)^{2m+\frac{1}{2}}}{m! \Gamma\left(m + \frac{3}{2}\right)} = \sqrt{\frac{2}{x\pi}} \sinh(x)$$

$$\sum_{n=0}^{\infty} B_n(x) \frac{z^n}{n!} = \frac{t}{e^t - 1} e^{tx} = \frac{te^{tx} e^{-\frac{t}{2}}}{e^{\frac{t}{2}} - e^{-\frac{t}{2}}} = \frac{te^{t(x-\frac{1}{2})}}{2} \sinh\left(\frac{t}{2}\right)$$

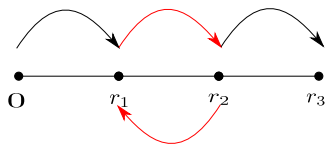
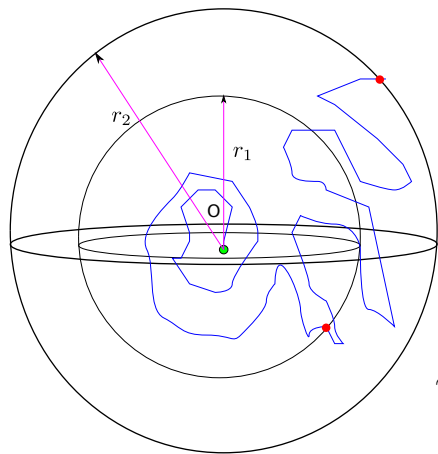
$$K_{\frac{1}{2}}(x) = \sqrt{\frac{\pi}{2x}} e^{-x}$$

$$\mathbb{E}_x \left(e^{-\alpha H_z}; \sup_{0 \leq s \leq H_z} R_s^{(3)} < y \right) = \begin{cases} \frac{z \sinh(x\sqrt{2\alpha})}{x \sinh(z\sqrt{2\alpha})}, & 0 \leq x \leq z \leq y \\ \frac{z \sinh((y-x)\sqrt{2\alpha})}{x \sinh((y-z)\sqrt{2\alpha})}, & z \leq x \leq y \end{cases}$$

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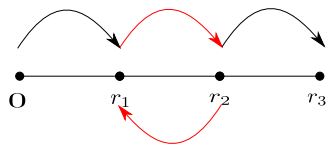
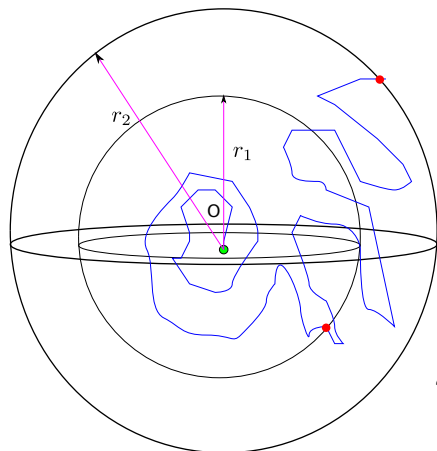
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$$0 \rightarrow r_3$$

$$\sim 0 \rightarrow r_1 + \boxed{r_1 \leftrightarrow r_2} + r_2 \rightarrow r_3$$

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$$r_1 = 1, r_2 = 2, \text{ and } r_3 = 3$$

$$\frac{3^n}{n+1} \left[B_{n+1} \left(\frac{x+5}{6} \right) - B_{n+1} \left(\frac{x+3}{6} \right) \right] = \frac{1}{4} \sum_{k=0}^{\infty} \frac{1}{4^k} E_n^{(2k+2)} \left(\frac{x+2k+3}{2} \right).$$

$$\sum_{n=0}^{\infty} E_n(x) \frac{z^n}{n!} = \frac{2}{e^z + 1} e^{xz} : E_n(x) = \int_{\mathbb{R}} \left(x + it - \frac{1}{2}\right)^n \operatorname{sech}(\pi t) dt.$$

$$\sum_{n=0}^{\infty} B_n(x) \frac{z^n}{n!} = \frac{t}{e^t - 1} e^{tx} : B_n(x) = \frac{\pi}{2} \int_{\mathbb{R}} \left(x + it - \frac{1}{2}\right)^n \operatorname{sech}^2(\pi t) dt.$$