

Tutorial on Additive Lévy Processes

Lecture #1

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Some Terminology

- An “ (N, d) random field” X has N parameters and takes values in \mathbf{R}^d (Adler, 1981);
i.e., $X(\mathbf{t}) \in \mathbf{R}^d$ for all $\mathbf{t} := (t_1, \dots, t_N) \in \mathbf{R}^N$.



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- X_1, \dots, X_N independent Brownian motions in \mathbf{R}^d .
“Additive Brownian motion”:

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- Likewise, can have “additive stable,” “additive Lévy,” etc.



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Lemma

If A_1, A_2, \dots are nonrandom and disjoint then a.s.,

$$\dot{W}\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} \dot{W}(A_n),$$

as long as $\text{meas}(A_n) < \infty$ for all n .



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- $(N, 1)$ “Brownian sheet” $B :=$ the dF of \dot{W} ; i.e.,

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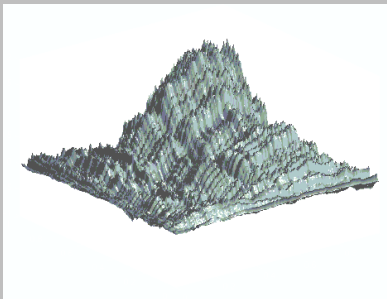
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- B is the noise in many SPDEs.
- In many cases, “SPDEs look like B .” [Girsanov]
- (N, d) Brownian sheet $B(\mathbf{t}) := (B_1(\mathbf{t}), \dots, B_d(\mathbf{t}))$, where B_1, \dots, B_d are indept. $(N, 1)$ Brownian sheets.



ABM and the Local Dynamics of Brownian Sheet



$(2, 1)$ Brownian sheet



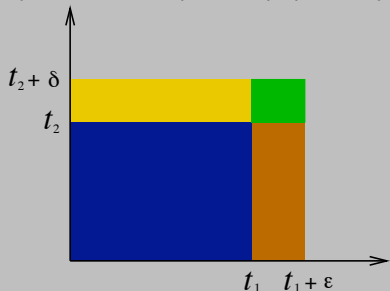
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- Let $B(t_1, t_2)$ denote 2-parameter Brownian sheet in \mathbf{R}^d .



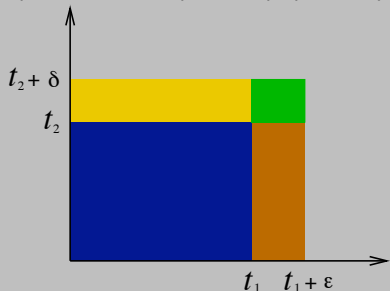
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- Let $B(t_1, t_2)$ denote 2-parameter Brownian sheet in \mathbf{R}^d .
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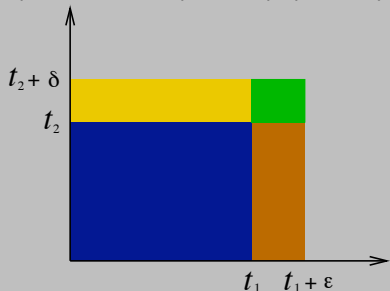


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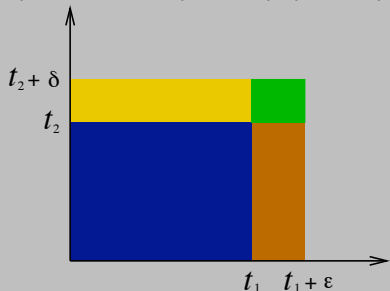


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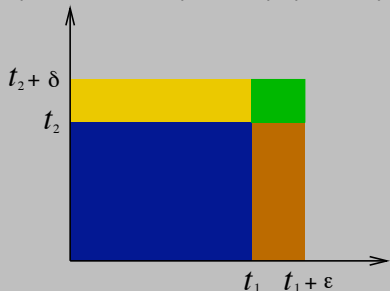


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$$B(1 + \epsilon, 1 + \delta) - \underbrace{B(1, 1)}^{\dot{W}(\blacksquare)} = \underbrace{X(\epsilon) + Y(\delta)}^{\dot{W}(\blacksquare) + \dot{W}(\blacksquare) = \text{ABM}} + \underbrace{Z(\epsilon, \delta)}^{\dot{W}(\blacksquare) = \text{BS}}$$



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References: Orey and Pruitt (1973), Kendall (1980), Ehm (1981), Dalang and Walsh (1992; 1993; 1996), Kh. (1995; 1999; 2003), Dalang and Mountford (1996; 1997; 2001), Kh. and Shi (1999), Kh. Xiao (2005), Kh. Xiao, and Wu (2006)



First Application: Contours of Brownian Sheet

Kendall's Theorem

Let $B = (2, 1)$ BS; choose and fix $s, t > 0$.



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- WLOG $s = t = 1$.



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- $C(r) := \{(x, y) \in \mathbf{R}^2 : |x - 1| \vee |y - 1| \leq r\}.$



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- I will sketch the proof of a slightly weaker variant. [In fact it is equivalent.]



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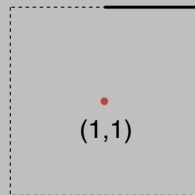
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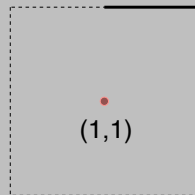
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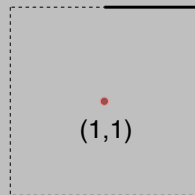
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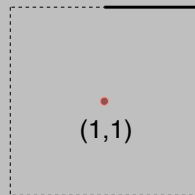


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- Goal: $\lim_{r \downarrow 0} P(J'(r)) > 0$.
- Local dynamics + scaling \Rightarrow

$$P(J'(r)) \rightarrow P \left\{ X(1) + Y(1) > \sup_{(u,v) \in C'(1)} (X(u) + Y(v)) \right\} > 0.$$



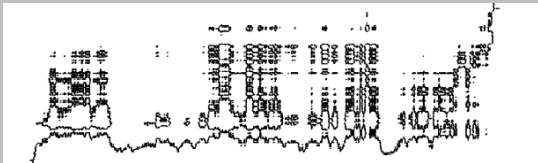
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Kendall's theorem holds for "most" points in the zero-set too.
Can you see it?



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(Kh., Révész, and Shi, 2005)



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem

X and $Y =$ independent BMs in \mathbf{R}^d . Then:
“Two BM paths can cross only in $\dim \leq 3$.”



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- 2 $d \leq 3$.



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The Dvoretzky–Erdős–Kakutani Theorem, Continued

Equivalently, TFAE:

$$X([1, 2]) \cap Y([1, 2]) \neq \emptyset \iff d \leq 3. \quad (1)$$



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Key idea: $X([1, 2]) \cap Y([1, 2]) \neq \emptyset \Leftrightarrow$

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The Dvoretzky–Erdős–Kakutani Theorem, Continued

- When $d \leq 3$ one can directly construct a “local time.” This is an a.s.-nontrivial random measure on the set $A^{-1}\{0\} = \{(s, t) \in [1, 2]^2 : X(s) = Y(t)\}$. Therefore, $0 \in A([1, 2]^2)$ a.s.



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- We prove that if $d \geq 5$ then $A(s, t) = X(s) - Y(t) \neq 0$ for all $s, t \in [1, 2]$. This proof can be pushed through when $d = 4$ but needs more care; see the original paper (Kh., *Expos. Math.*, 2003).



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Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

- When $d \leq 3$ one can directly construct a “local time.” This is an a.s.-nontrivial random measure on the set $A^{-1}\{0\} = \{(s, t) \in [1, 2]^2 : X(s) = Y(t)\}$. Therefore, $0 \in A([1, 2]^2)$ a.s.
- We prove that if $d \geq 5$ then $A(s, t) = X(s) - Y(t) \neq 0$ for all $s, t \in [1, 2]$. This proof can be pushed through when $d = 4$ but needs more care; see the original paper (Kh., *Expos. Math.*, 2003).

Exercise

Prove that $A([1, 2]^2) = A(1, 1) + \bar{A}([0, 1]^2)$, where \bar{A} is a copy of A , independent of $A(1, 1)$, and $a + S = \{a + s : s \in S\}$ for all $a \in \mathbf{R}^d$ and $S \subset \mathbf{R}^d$.

Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.
By the Exercise,

$$P \left\{ 0 \in A \left([1, 2]^2 \right) \right\}$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

By the Exercise,

$$P\{0 \in A([1, 2]^2)\} = \int_{\mathbf{R}^d} P\{x \in A([0, 1]^2)\} \underbrace{P\{A(1, 1) \in -dx\}}$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

By the Exercise,

$$P\{0 \in A([1, 2]^2)\} = \int_{\mathbf{R}^d} P\{x \in A([0, 1]^2)\} \underbrace{P\{A(1, 1) \in -dx\}}_{=\varphi(x)dx, \varphi > 0}$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

By the Exercise,

$$\begin{aligned}
 P\{0 \in A([1, 2]^2)\} &= \int_{\mathbf{R}^d} P\{x \in A([0, 1]^2)\} \underbrace{P\{A(1, 1) \in -dx\}}_{=\varphi(x)dx, \varphi > 0} \\
 &\asymp \int_{\mathbf{R}^d} P\{x \in A([0, 1]^2)\} dx
 \end{aligned}$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

By the Exercise,

$$\begin{aligned}
 P\{0 \in A([1, 2]^2)\} &= \int_{\mathbf{R}^d} P\{x \in A([0, 1]^2)\} \underbrace{P\{A(1, 1) \in -dx\}}_{=\varphi(x)dx, \varphi > 0} \\
 &\asymp \int_{\mathbf{R}^d} P\{x \in A([0, 1]^2)\} dx \\
 &= E\{m(A([0, 1]^2))\}.
 \end{aligned}$$

(Lévy, 1940; Kahane, 1983) “ $a \asymp b$ ” means “ $a > 0 \Leftrightarrow b > 0$ ”.



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

Goal: $Em(A([0, 1]^2)) = 0$.



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

Goal: $Em(A([0, 1]^2)) = 0$.

By Brownian scaling,

$$\underbrace{A([0, 2]^2)}$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

Goal: $Em(A([0, 1]^2)) = 0$.

By Brownian scaling,

$$\underbrace{A([0, 2]^2)}_{\sim X(2s) - Y(2t)}$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

Goal: $Em(A([0, 1]^2)) = 0$.

By Brownian scaling,

$$\underbrace{A([0, 2]^2)}_{\sim X(2s) - Y(2t)} \stackrel{\mathcal{D}}{=} \sqrt{2}A([0, 1]^2)$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

Goal: $Em(A([0, 1]^2)) = 0$.

By Brownian scaling,

$$\underbrace{A([0, 2]^2)}_{\sim X(2s) - Y(2t)} \stackrel{\mathcal{D}}{=} \sqrt{2}A([0, 1]^2)$$

$$\Leftrightarrow m(A([0, 2]^2)) \stackrel{\mathcal{D}}{=} 2^{d/2}m(A[0, 1]^2)$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

Goal: $Em(A([0, 1]^2)) = 0$.

By Brownian scaling,

$$\underbrace{A([0, 2]^2)}_{\sim X(2s) - Y(2t)} \stackrel{\mathcal{D}}{=} \sqrt{2}A([0, 1]^2)$$

$$\begin{aligned} \Leftrightarrow m(A([0, 2]^2)) &\stackrel{\mathcal{D}}{=} 2^{d/2}m(A[0, 1]^2) \\ \Leftrightarrow Em(A([0, 2]^2)) &= 2^{d/2}Em(A([0, 1]^2)). \end{aligned}$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

We know: $Em(A([0, 2]^2)) = 2^{d/2}Em(A([0, 1]^2))$.



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

We know: $Em(A([0, 2]^2)) = 2^{d/2}Em(A([0, 1]^2))$.

We want: $Em(A([0, 1]^2)) = 0$.



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

We know: $Em(A([0, 2]^2)) = 2^{d/2}Em(A([0, 1]^2))$.

We want: $Em(A([0, 1]^2)) = 0$.

Note that

$$\begin{aligned} A([0, 2]^2) &= A([0, 1]^2) \cup A([0, 1] \times [1, 2]) \\ &\quad \cup A([1, 2] \times [0, 1]) \cup A([1, 2]^2). \end{aligned}$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Let $A(s, t) := X(s) - Y(t)$, $d \geq 5$; $m := \text{Leb. meas. on } \mathbf{R}^d$.

We know: $Em(A([0, 2]^2)) = 2^{d/2}Em(A([0, 1]^2))$.

We want: $Em(A([0, 1]^2)) = 0$.

Note that

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Therefore,

$$\begin{aligned} m(A([0, 2]^2)) &\leq m(A([0, 1]^2)) + m(A([0, 1] \times [1, 2])) \\ &\quad + m(A([1, 2] \times [0, 1])) + m(A([1, 2]^2)). \end{aligned}$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

We know:

$$\begin{aligned} m\left(A\left([0, 2]^2\right)\right) &\leq m\left(A\left([0, 1]^2\right)\right) + m\left(A\left([0, 1] \times [1, 2]\right)\right) \\ &\quad + m\left(A\left([1, 2] \times [0, 1]\right)\right) + m\left(A\left([1, 2]^2\right)\right). \end{aligned}$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

We know:

$$m\left(A\left([0, 2]^2\right)\right) \leq m\left(A\left([0, 1]^2\right)\right) + m\left(A\left([0, 1] \times [1, 2]\right)\right) \\ + m\left(A\left([1, 2] \times [0, 1]\right)\right) + m\left(A\left([1, 2]^2\right)\right).$$

By the Exercise,

$$\Rightarrow m\left(A\left([1, 2]^2\right)\right) \stackrel{\mathcal{D}}{=} m\left(A\left([0, 1]^2\right)\right).$$

Same with the other terms in the first display.



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Thus,

$$Em\left(A\left([0, 2]^2\right)\right) \leq 4Em\left(A\left([0, 1]^2\right)\right)$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Thus,

$$Em\left(A\left([0, 2]^2\right)\right) \leq 4Em\left(A\left([0, 1]^2\right)\right)$$

\Rightarrow

$$2^{d/2}Em\left(A\left([0, 1]^2\right)\right)$$



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Thus,

$$Em\left(A\left([0, 2]^2\right)\right) \leq 4Em\left(A\left([0, 1]^2\right)\right)$$

\Rightarrow

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Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Thus,

$$Em\left(A\left([0, 2]^2\right)\right) \leq 4Em\left(A\left([0, 1]^2\right)\right)$$

\Rightarrow

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$d \geq 5 \Rightarrow Em(A([0, 1]^2)) = 0$, as desired.



Second Application: Intersections of Brownian Motions

The Dvoretzky–Erdős–Kakutani Theorem, Continued

Thus,

$$Em\left(A\left([0, 2]^2\right)\right) \leq 4Em\left(A\left([0, 1]^2\right)\right)$$

\Rightarrow

$$2^{d/2}Em\left(A\left([0, 1]^2\right)\right) \leq 4Em\left(A\left([0, 1]^2\right)\right).$$

$d \geq 5 \Rightarrow Em(A([0, 1]^2)) = 0$, as desired.

Exercise

Prove that $Em(A([0, 1]^2)) < \infty$.



In Memory of Ron Pyke (1931–2005)

