

# Lecture 2

## The Range of a Lévy Process

Davar Khoshnevisan

Department of Mathematics  
University of Utah

<http://www.math.utah.edu/~davar>

Summer School on Lévy Processes: Theory and Applications  
August 9–12, 2007  
Sandbjerg Manor, Denmark



# Recap

1. Suppose  $\exists C > 0$  such that  $\forall \epsilon > 0 \exists$  and  $b$ -adic cubes  $F_1, F_2, \dots$  with  $\text{diam} F_j \leq \epsilon$  and  $F \subseteq \bigcup_{j=1}^{\infty} F_j$  such that  $\sum_{j=1}^{\infty} |\text{diam} F_j|^s \leq C$ . Then  $\dim_{\text{H}} F \leq s$ .



# Recap

1. Suppose  $\exists C > 0$  such that  $\forall \epsilon > 0 \exists$  and  $b$ -adic cubes  $F_1, F_2, \dots$  with  $\text{diam} F_j \leq \epsilon$  and  $F \subseteq \bigcup_{j=1}^{\infty} F_j$  such that  $\sum_{j=1}^{\infty} |\text{diam} F_j|^s \leq C$ . Then  $\dim_{\text{H}} F \leq s$ .
2. Suppose  $\exists \mu \in \mathcal{P}(F)$  such that  $I_s(\mu) < \infty$ , where

$$I_s(\mu) := \iint \frac{\mu(dx) \mu(dy)}{|x - y|^s}.$$

Then  $\dim_{\text{H}} F \geq s$ .



# Goal

- ▶ Find a formula for  $\dim_{\text{H}} X([0, 1])$ ,  $\dim_{\text{H}} X(\mathbf{R}_+)$ ,  $\overline{\dim}_{\text{M}} X([0, 1])$ , etc., where  $X$  is a Lévy process.



# Goal

- ▶ Find a formula for  $\dim_{\text{H}} X([0, 1])$ ,  $\dim_{\text{H}} X(\mathbf{R}_+)$ ,  $\overline{\dim}_{\text{M}} X([0, 1])$ , etc., where  $X$  is a Lévy process.
- ▶ Two issues:



# Goal

- ▶ Find a formula for  $\dim_{\mathbf{H}} X([0, 1])$ ,  $\dim_{\mathbf{H}} X(\mathbf{R}_+)$ ,  $\overline{\dim}_{\mathbf{M}} X([0, 1])$ , etc., where  $X$  is a Lévy process.
- ▶ Two issues:
  1. A lower bound [Uses an abstract form of the Frostman theorem]



# Goal

- ▶ Find a formula for  $\dim_{\text{H}} X([0, 1])$ ,  $\dim_{\text{H}} X(\mathbf{R}_+)$ ,  $\overline{\dim}_{\text{M}} X([0, 1])$ , etc., where  $X$  is a Lévy process.
- ▶ Two issues:
  1. A lower bound [Uses an abstract form of the Frostman theorem]
  2. An upper bound [Most Lévy processes are not continuous; the method for BM fails.]



# Goal

- ▶ Find a formula for  $\dim_{\mathbf{H}} X([0, 1])$ ,  $\dim_{\mathbf{H}} X(\mathbf{R}_+)$ ,  $\overline{\dim}_{\mathbf{M}} X([0, 1])$ , etc., where  $X$  is a Lévy process.
- ▶ Two issues:
  1. A lower bound [Uses an abstract form of the Frostman theorem]
  2. An upper bound [Most Lévy processes are not continuous; the method for BM fails.]
  3. We will handle these matters in reverse order.





# Potential measures

- ▶  $X :=$  a Lévy process in  $\mathbf{R}^d$ .



# Potential measures

- ▶  $X$  := a Lévy process in  $\mathbf{R}^d$ .
- ▶ For all Borel sets  $A \subseteq \mathbf{R}^d$  define

$$U(A) := \int_0^\infty \mathbf{P}\{X(s) \in A\} e^{-s} ds.$$

[The one-potential measure]



# Potential measures

- ▶  $X$  := a Lévy process in  $\mathbf{R}^d$ .
- ▶ For all Borel sets  $A \subseteq \mathbf{R}^d$  define

$$U(A) := \int_0^\infty \mathbf{P}\{X(s) \in A\} e^{-s} ds.$$

[The one-potential measure]

- ▶  $\zeta$  := independent mean-one exponential.



# Potential measures

- ▶  $X$  := a Lévy process in  $\mathbf{R}^d$ .
- ▶ For all Borel sets  $A \subseteq \mathbf{R}^d$  define

$$U(A) := \int_0^\infty \mathbf{P}\{X(s) \in A\} e^{-s} ds.$$

[The one-potential measure]

- ▶  $\zeta$  := independent mean-one exponential.
- ▶  $\mathbf{P}\{\zeta > s\} = \exp(-s)$ .



# Potential measures

- ▶  $X :=$  a Lévy process in  $\mathbf{R}^d$ .
- ▶ For all Borel sets  $A \subseteq \mathbf{R}^d$  define

$$U(A) := \int_0^\infty \mathbf{P}\{X(s) \in A\} e^{-s} ds.$$

[The one-potential measure]

- ▶  $\zeta :=$  independent mean-one exponential.
- ▶  $\mathbf{P}\{\zeta > s\} = \exp(-s)$ .
- ▶  $U(A) = \mathbf{E} \left[ \int_0^\zeta \mathbf{1}_A(X(s)) ds \right]$ .



# Potential measures

- ▶  $X :=$  a Lévy process in  $\mathbf{R}^d$ .
- ▶ For all Borel sets  $A \subseteq \mathbf{R}^d$  define

$$U(A) := \int_0^\infty \mathbf{P}\{X(s) \in A\} e^{-s} ds.$$

[The one-potential measure]

- ▶  $\zeta :=$  independent mean-one exponential.
- ▶  $\mathbf{P}\{\zeta > s\} = \exp(-s)$ .
- ▶  $U(A) = \mathbf{E} \left[ \int_0^\zeta \mathbf{1}_A(X(s)) ds \right]$ .
- ▶  $U$  is a Borel probability measure on  $\mathbf{R}^d$ .



# A hitting bound

For all  $a \in \mathbf{R}^d$  and  $\epsilon > 0$  define

$$B(a, \epsilon) := \bigcap_{j=1}^d \left\{ x \in \mathbf{R}^d : a_j - \epsilon \leq x_j < a_j + \epsilon \right\}.$$



# A hitting bound

For all  $a \in \mathbf{R}^d$  and  $\epsilon > 0$  define

$$B(a, \epsilon) := \bigcap_{j=1}^d \left\{ x \in \mathbf{R}^d : a_j - \epsilon \leq x_j < a_j + \epsilon \right\}.$$

## Lemma

For all  $a \in \mathbf{R}^d$  and  $\epsilon > 0$ ,

$$\mathbf{P} \{ X(s) \in B(a, \epsilon) \text{ for some } s \leq \zeta \} \leq \frac{U(B(a, 2\epsilon))}{U(B(0, \epsilon))}.$$





# A hitting bound

- ▶ **Proof:** Let  $T$  denote the first hitting time of  $B(a, \epsilon)$ .



## A hitting bound

- ▶ **Proof:** Let  $T$  denote the first hitting time of  $B(a, \epsilon)$ .
- ▶ By the strong Markov property,

$$\begin{aligned} & U(B(0, \epsilon))P\{\zeta > T\} \\ &= E\left(\int_0^\infty \mathbf{1}_{B(0, \epsilon)}(X(s+T) - X(T))e^{-(s+T)} ds; T < \infty\right), \end{aligned}$$

since  $P\{\zeta > T\} = E[e^{-T}; T < \infty]$ .



## A hitting bound

- ▶ **Proof:** Let  $T$  denote the first hitting time of  $B(a, \epsilon)$ .
- ▶ By the strong Markov property,

$$\begin{aligned} & U(B(0, \epsilon))P\{\zeta > T\} \\ &= E\left(\int_0^\infty \mathbf{1}_{B(0, \epsilon)}(X(s+T) - X(T))e^{-(s+T)} ds; T < \infty\right), \end{aligned}$$

since  $P\{\zeta > T\} = E[e^{-T}; T < \infty]$ .

- ▶  $|X(T) - a| \leq \epsilon$  a.s. on  $\{T < \infty\}$ .



## A hitting bound

- ▶ **Proof:** Let  $T$  denote the first hitting time of  $B(a, \epsilon)$ .
- ▶ By the strong Markov property,

$$\begin{aligned} & U(B(0, \epsilon))\mathbf{P}\{\zeta > T\} \\ &= \mathbf{E} \left( \int_0^\infty \mathbf{1}_{B(0, \epsilon)}(X(s+T) - X(T))e^{-(s+T)} ds; T < \infty \right), \end{aligned}$$

since  $\mathbf{P}\{\zeta > T\} = \mathbf{E}[e^{-T}; T < \infty]$ .

- ▶  $|X(T) - a| \leq \epsilon$  a.s. on  $\{T < \infty\}$ .
- ▶ By the triangle inequality,

$$\begin{aligned} & U(B(0, \epsilon))\mathbf{P}\{\zeta > T\} \\ &\leq \mathbf{E} \left( \int_0^\infty \mathbf{1}_{B(a, 2\epsilon)}(X(s+T))e^{-(s+T)} ds; T < \infty \right). \end{aligned}$$



# A hitting bound

By the triangle inequality,

$$\begin{aligned} & U(B(0, \epsilon))P\{\zeta > T\} \\ & \leq E \left( \int_0^\infty \mathbf{1}_{B(a, 2\epsilon)}(X(s+T))e^{-(s+T)} ds; T < \infty \right). \end{aligned}$$

Therefore,

$$\begin{aligned} & U(B(0, \epsilon))P\{\zeta > T\} \\ & \leq E \left( \int_T^\infty \mathbf{1}_{B(a, 2\epsilon)}(X(s))e^{-s} ds \right). \end{aligned}$$



# A hitting bound

By the triangle inequality,

$$\begin{aligned} & U(B(0, \epsilon))P\{\zeta > T\} \\ & \leq E \left( \int_0^\infty \mathbf{1}_{B(a, 2\epsilon)}(X(s+T))e^{-(s+T)} ds; T < \infty \right). \end{aligned}$$

Therefore,

$$\begin{aligned} & U(B(0, \epsilon))P\{\zeta > T\} \\ & \leq E \left( \int_T^\infty \mathbf{1}_{B(a, 2\epsilon)}(X(s))e^{-s} ds \right). \end{aligned}$$

This is  $\leq U(B(a, 2\epsilon))$ . ■



# A hitting bound

- ▶ Recall that  $N_n(X([0, t]))$  denotes the number of dyadic cubes of side  $2^{-n}$  the intersect  $X([0, t])$ .



# A hitting bound

- ▶ Recall that  $N_n(X([0, t]))$  denotes the number of dyadic cubes of side  $2^{-n}$  that intersect  $X([0, t])$ .
- ▶ By the lemma,

$$E(N_n(X([0, \zeta]))) \leq \frac{1}{U(B(0, 2^{-n}))} \sum_{B(a, 2^{-n}) \in \mathcal{D}_n} U(B(a, 2^{-n+1})),$$

where  $\mathcal{D}_n :=$  all dyadic cubes.





# A hitting bound

- ▶ Recall that  $N_n(X([0, t]))$  denotes the number of dyadic cubes of side  $2^{-n}$  that intersect  $X([0, t])$ .
- ▶ By the lemma,

$$E(N_n(X([0, \zeta]))) \leq \frac{1}{U(B(0, 2^{-n}))} \sum_{B(a, 2^{-n}) \in \mathcal{D}_n} U(B(a, 2^{-n+1})),$$

where  $\mathcal{D}_n :=$  all dyadic cubes.

- ▶ Now  $\sum_{B(a, 2^{-n}) \in \mathcal{D}_n} U(B(a, 2^{-n})) = 1$ .



# A hitting bound

- ▶ Recall that  $N_n(X([0, t]))$  denotes the number of dyadic cubes of side  $2^{-n}$  that intersect  $X([0, t])$ .
- ▶ By the lemma,

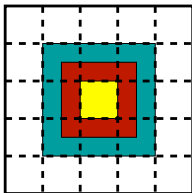
$$E(N_n(X([0, \zeta]))) \leq \frac{1}{U(B(0, 2^{-n}))} \sum_{B(a, 2^{-n}) \in \mathcal{D}_n} U(B(a, 2^{-n+1})),$$

where  $\mathcal{D}_n :=$  all dyadic cubes.

- ▶ Now  $\sum_{B(a, 2^{-n}) \in \mathcal{D}_n} U(B(a, 2^{-n})) = 1$ .
- ▶ What about  $\sum_{B(a, 2^{-n}) \in \mathcal{D}_n} U(B(a, 2^{-n+1}))$ ?



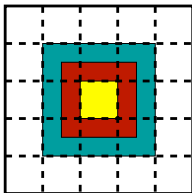
# A hitting bound



- ▶ Yellow =  $B(a, 2^{-n})$ ; a dyadic ball.



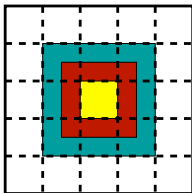
# A hitting bound



- ▶ Yellow =  $B(a, 2^{-n})$ ; a dyadic ball.
- ▶ Red =  $B(a, 2^{-n+1})$ .



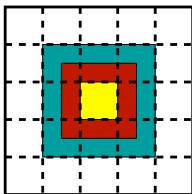
# A hitting bound



- ▶ Yellow =  $B(a, 2^{-n})$ ; a dyadic ball.
- ▶ Red =  $B(a, 2^{-n+1})$ .
- ▶ Blue + yellow =  $3^d$  dyadic balls of radius  $2^{-n}$  each.  
*Clique*



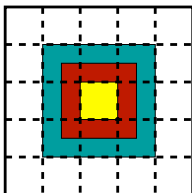
# A hitting bound



- ▶ Yellow =  $B(a, 2^{-n})$ ; a dyadic ball.
- ▶ Red =  $B(a, 2^{-n+1})$ .
- ▶ Blue + yellow =  $3^d$  dyadic balls of radius  $2^{-n}$  each.  
*Clique*
- ▶ Each  $I \in \mathcal{D}_n$  is in at most  $5^d$  cliques.



# A hitting bound



- ▶ Yellow =  $B(a, 2^{-n})$ ; a dyadic ball.
- ▶ Red =  $B(a, 2^{-n+1})$ .
- ▶ Blue + yellow =  $3^d$  dyadic balls of radius  $2^{-n}$  each.  
*Clique*
- ▶ Each  $I \in \mathcal{D}_n$  is in at most  $5^d$  cliques.
- ▶  $\sum_{B(a, 2^{-n}) \in \mathcal{D}_n} U(B(a, 2^{-n+1})) \leq 5^d \sum_{I \in \mathcal{D}_n} U(I) = 5^d$ .



# A hitting bound

► Thus,

$$E(N_n(X([0, \zeta]))) \leq \frac{5^d}{U(B(0, 2^{-n}))}.$$





# A hitting bound

- ▶ Thus,

$$E(N_n(X([0, \zeta]))) \leq \frac{5^d}{U(B(0, 2^{-n}))}.$$

- ▶  $N_n(X([0, \zeta])) \geq N_n(X([0, 1]))$  on  $\{\zeta > 1\}$ .



# A hitting bound

- ▶ Thus,

$$E(N_n(X([0, \zeta]))) \leq \frac{5^d}{U(B(0, 2^{-n}))}.$$

- ▶  $N_n(X([0, \zeta])) \geq N_n(X([0, 1]))$  on  $\{\zeta > 1\}$ .
- ▶  $\Rightarrow E(N_n(X([0, \zeta]))) \geq e^{-1}E(N_n(X([0, 1])))$ .



# A hitting bound

- ▶ Thus,

$$E(N_n(X([0, \zeta]))) \leq \frac{5^d}{U(B(0, 2^{-n}))}.$$

- ▶  $N_n(X([0, \zeta])) \geq N_n(X([0, 1]))$  on  $\{\zeta > 1\}$ .
- ▶  $\Rightarrow E(N_n(X([0, \zeta]))) \geq e^{-1}E(N_n(X([0, 1])))$ .

## Lemma

$$E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})).$$



# A hitting bound

- ▶ Thus,

$$E(N_n(X([0, \zeta]))) \leq \frac{5^d}{U(B(0, 2^{-n}))}.$$

- ▶  $N_n(X([0, \zeta])) \geq N_n(X([0, 1]))$  on  $\{\zeta > 1\}$ .
- ▶  $\Rightarrow E(N_n(X([0, \zeta]))) \geq e^{-1}E(N_n(X([0, 1])))$ .

## Lemma

$$E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})).$$



# A hitting bound

- ▶ Thus,

$$E(N_n(X([0, \zeta]))) \leq \frac{5^d}{U(B(0, 2^{-n}))}.$$

- ▶  $N_n(X([0, \zeta])) \geq N_n(X([0, 1]))$  on  $\{\zeta > 1\}$ .
- ▶  $\Rightarrow E(N_n(X([0, \zeta]))) \geq e^{-1}E(N_n(X([0, 1])))$ .

## Lemma

$$E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})).$$

This will give an upper bound for  $\dim_{\text{H}} X([0, 1])$ , and another upper bound for  $\dim_{\text{M}} X([0, 1])$ .



# Potential indices

- ▶  $\overline{\text{ind}}U := \limsup_{\epsilon \rightarrow 0} \log U(B(0, \epsilon)) / \log \epsilon.$   
[Upper index]



# Potential indices

- ▶  $\overline{\text{ind}}U := \limsup_{\epsilon \rightarrow 0} \log U(B(0, \epsilon)) / \log \epsilon.$   
[Upper index]
- ▶  $\underline{\text{ind}}U := \liminf_{\epsilon \rightarrow 0} \log U(B(0, \epsilon)) / \log \epsilon.$   
[Lower index]



# Potential indices

- ▶  $\overline{\text{ind}}U := \limsup_{\epsilon \rightarrow 0} \log U(B(0, \epsilon)) / \log \epsilon.$   
[Upper index]
- ▶  $\underline{\text{ind}}U := \liminf_{\epsilon \rightarrow 0} \log U(B(0, \epsilon)) / \log \epsilon.$   
[Lower index]
- ▶  $U(B(0, \epsilon)) \geq \epsilon^{\underline{\text{ind}}U + o(1)}$  for infinitely many  $\epsilon$  small.





# Potential indices

- ▶  $\overline{\text{ind}}U := \limsup_{\epsilon \rightarrow 0} \log U(B(0, \epsilon)) / \log \epsilon$ .  
[Upper index]
- ▶  $\underline{\text{ind}}U := \liminf_{\epsilon \rightarrow 0} \log U(B(0, \epsilon)) / \log \epsilon$ .  
[Lower index]
- ▶  $U(B(0, \epsilon)) \geq \epsilon^{\underline{\text{ind}}U + o(1)}$  for infinitely many  $\epsilon$  small.
- ▶  $U(B(0, \epsilon)) \geq \epsilon^{\overline{\text{ind}}U + o(1)}$  for all  $\epsilon$  small.



# Potential indices

- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \overline{\text{ind}} U + o(1)}$ , eventually.



# Potential indices

- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \overline{\text{ind}} U + o(1)}$ , eventually.
  1.  $s > \overline{\text{ind}} U \Rightarrow \sum_n P\{N_n(X([0, 1])) \geq 2^{ns}\} < \infty$ .



# Potential indices

- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \overline{\text{ind}} U + o(1)}$ , eventually.
  1.  $s > \overline{\text{ind}} U \Rightarrow \sum_n P\{N_n(X([0, 1])) \geq 2^{ns}\} < \infty$ .
  2.  $\overline{\text{dim}}_M X([0, 1]) \leq \overline{\text{ind}} U$  a.s.



# Potential indices

- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \overline{\text{ind}} U + o(1)}$ , eventually.
  1.  $s > \overline{\text{ind}} U \Rightarrow \sum_n P\{N_n(X([0, 1])) \geq 2^{ns}\} < \infty$ .
  2.  $\overline{\text{dim}}_M X([0, 1]) \leq \overline{\text{ind}} U$  a.s.
- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \underline{\text{ind}} U + o(1)}$ , i.o.



# Potential indices

- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \overline{\text{ind}} U + o(1)}$ , eventually.
  1.  $s > \overline{\text{ind}} U \Rightarrow \sum_n P\{N_n(X([0, 1])) \geq 2^{ns}\} < \infty$ .
  2.  $\overline{\text{dim}}_M X([0, 1]) \leq \overline{\text{ind}} U$  a.s.
- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \underline{\text{ind}} U + o(1)}$ , i.o.
  1.  $s > \underline{\text{ind}} U \Rightarrow N_n(X([0, 1])) \leq 2^{ns}$  i.o., a.s.



# Potential indices

- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \overline{\text{ind}} U + o(1)}$ , eventually.
  1.  $s > \overline{\text{ind}} U \Rightarrow \sum_n P\{N_n(X([0, 1])) \geq 2^{ns}\} < \infty$ .
  2.  $\overline{\text{dim}}_M X([0, 1]) \leq \overline{\text{ind}} U$  a.s.
- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \underline{\text{ind}} U + o(1)}$ , i.o.
  1.  $s > \underline{\text{ind}} U \Rightarrow N_n(X([0, 1])) \leq 2^{ns}$  i.o., a.s.
  2.  $\underline{\text{dim}}_H X([0, 1]) \leq \underline{\text{ind}} U$  a.s.



# Potential indices

- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \overline{\text{ind}} U + o(1)}$ , eventually.
  1.  $s > \overline{\text{ind}} U \Rightarrow \sum_n P\{N_n(X([0, 1])) \geq 2^{ns}\} < \infty$ .
  2.  $\overline{\text{dim}}_M X([0, 1]) \leq \overline{\text{ind}} U$  a.s.
- ▶  $E[N_n(X([0, 1]))] \leq 5^d e / U(B(0, 2^{-n})) \leq 2^{n \underline{\text{ind}} U + o(1)}$ , i.o.
  1.  $s > \underline{\text{ind}} U \Rightarrow N_n(X([0, 1])) \leq 2^{ns}$  i.o., a.s.
  2.  $\text{dim}_H X([0, 1]) \leq \underline{\text{ind}} U$  a.s.
- ▶ Both bounds are sharp.  
[ $\text{dim}_H$  Pruitt, 1969;  $\overline{\text{dim}}_M$  Taylor, XXXX]





# Theorems of Pruitt and Taylor

## Theorem

A.s.:  $\dim_{\mathbb{H}} X([0, 1]) = \underline{\text{ind}}U$  and  $\overline{\dim}_{\mathbb{M}} X([0, 1]) = \overline{\text{ind}}U$ .

Derivation of the formula for  $\dim_{\mathbb{H}}$ :



# Theorems of Pruitt and Taylor

## Theorem

A.s.:  $\dim_{\mathbb{H}} X([0, 1]) = \underline{\text{ind}}U$  and  $\overline{\dim}_{\mathbb{M}} X([0, 1]) = \overline{\text{ind}}U$ .

## Derivation of the formula for $\dim_{\mathbb{H}}$ :

- ▶ For  $\dim_{\mathbb{H}}$ : Enough to derive  $\dim_{\mathbb{H}} X([0, 1]) \geq \underline{\text{ind}}U$ .



# Theorems of Pruitt and Taylor

## Theorem

A.s.:  $\dim_{\mathbb{H}} X([0, 1]) = \underline{\text{ind}}U$  and  $\overline{\dim}_{\mathbb{M}} X([0, 1]) = \overline{\text{ind}}U$ .

## Derivation of the formula for $\dim_{\mathbb{H}}$ :

- ▶ For  $\dim_{\mathbb{H}}$ : Enough to derive  $\dim_{\mathbb{H}} X([0, 1]) \geq \underline{\text{ind}}U$ .
- ▶ Let  $\mu(A) := \int_0^1 \mathbf{1}_A(X(s)) ds$  [Occupation measure]



# Theorems of Pruitt and Taylor

## Theorem

A.s.:  $\dim_{\mathbb{H}} X([0, 1]) = \underline{\text{ind}}U$  and  $\overline{\dim}_{\mathbb{M}} X([0, 1]) = \overline{\text{ind}}U$ .

## Derivation of the formula for $\dim_{\mathbb{H}}$ :

- ▶ For  $\dim_{\mathbb{H}}$ : Enough to derive  $\dim_{\mathbb{H}} X([0, 1]) \geq \underline{\text{ind}}U$ .
- ▶ Let  $\mu(A) := \int_0^1 \mathbf{1}_A(X(s)) ds$  [Occupation measure]
- ▶ Strategy:  $I_s(\mu) = \iint |x - y|^{-s} \mu(dx) \mu(dy) < \infty$ .



# Theorems of Pruitt and Taylor

▶  $I_s(\mu) = \int_0^1 \int_0^1 |X(u) - X(v)|^{-s} du dv.$



# Theorems of Pruitt and Taylor

- ▶  $I_s(\mu) = \int_0^1 \int_0^1 |X(u) - X(v)|^{-s} du dv.$
- ▶  $E[I_s(\mu)] = 2 \int_0^1 E(|X(u)|^{-s}) du.$



# Theorems of Pruitt and Taylor

- ▶  $I_s(\mu) = \int_0^1 \int_0^1 |X(u) - X(v)|^{-s} du dv.$
- ▶  $E[I_s(\mu)] = 2 \int_0^1 E(|X(u)|^{-s}) du.$
- ▶ If  $s < \underline{\text{ind}}U$ , then  $\int_0^1 P\{|X(u)| \leq \epsilon\} du = O(\epsilon^s)$ . Therefore, for all  $0 < s < t < \underline{\text{ind}}U$ ,

$$\int_0^1 E(|X(u)|^{-s}) du = \int_0^1 \int_0^\infty P\{|X(u)|^{-s} > \lambda\} d\lambda du$$



# Theorems of Pruitt and Taylor

- ▶  $I_s(\mu) = \int_0^1 \int_0^1 |X(u) - X(v)|^{-s} du dv.$
- ▶  $E[I_s(\mu)] = 2 \int_0^1 E(|X(u)|^{-s}) du.$
- ▶ If  $s < \underline{\text{ind}}U$ , then  $\int_0^1 P\{|X(u)| \leq \epsilon\} du = O(\epsilon^s)$ . Therefore, for all  $0 < s < t < \underline{\text{ind}}U$ ,

$$\int_0^1 E(|X(u)|^{-s}) du = \int_0^1 \int_0^\infty P\{|X(u)|^{-s} > \lambda\} d\lambda du$$





# Theorems of Pruitt and Taylor

- ▶  $I_s(\mu) = \int_0^1 \int_0^1 |X(u) - X(v)|^{-s} du dv.$
- ▶  $E[I_s(\mu)] = 2 \int_0^1 E(|X(u)|^{-s}) du.$
- ▶ If  $s < \underline{\text{ind}}U$ , then  $\int_0^1 P\{|X(u)| \leq \epsilon\} du = O(\epsilon^s)$ . Therefore, for all  $0 < s < t < \underline{\text{ind}}U$ ,

$$\begin{aligned} \int_0^1 E(|X(u)|^{-s}) du &= \int_0^1 \int_0^\infty P\{|X(u)|^{-s} > \lambda\} d\lambda du \\ &\leq 1 + \int_1^\infty \int_0^1 P\{|X(u)| \leq \lambda^{-1/s}\} du d\lambda \end{aligned}$$



# Theorems of Pruitt and Taylor

- ▶  $I_s(\mu) = \int_0^1 \int_0^1 |X(u) - X(v)|^{-s} du dv.$
- ▶  $E[I_s(\mu)] = 2 \int_0^1 E(|X(u)|^{-s}) du.$
- ▶ If  $s < \underline{\text{ind}}U$ , then  $\int_0^1 P\{|X(u)| \leq \epsilon\} du = O(\epsilon^s)$ . Therefore, for all  $0 < s < t < \underline{\text{ind}}U$ ,

$$\begin{aligned} \int_0^1 E(|X(u)|^{-s}) du &= \int_0^1 \int_0^\infty P\{|X(u)|^{-s} > \lambda\} d\lambda du \\ &\leq 1 + \int_1^\infty \int_0^1 P\{|X(u)| \leq \lambda^{-1/s}\} du d\lambda \\ &\leq 1 + C \int_1^\infty \lambda^{-t/s} d\lambda < \infty. \end{aligned}$$



# Theorems of Pruitt and Taylor

- ▶  $I_s(\mu) = \int_0^1 \int_0^1 |X(u) - X(v)|^{-s} du dv.$
- ▶  $E[I_s(\mu)] = 2 \int_0^1 E(|X(u)|^{-s}) du.$
- ▶ If  $s < \underline{\text{ind}}U$ , then  $\int_0^1 P\{|X(u)| \leq \epsilon\} du = O(\epsilon^s)$ . Therefore, for all  $0 < s < t < \underline{\text{ind}}U$ ,

$$\begin{aligned} \int_0^1 E(|X(u)|^{-s}) du &= \int_0^1 \int_0^\infty P\{|X(u)|^{-s} > \lambda\} d\lambda du \\ &\leq 1 + \int_1^\infty \int_0^1 P\{|X(u)| \leq \lambda^{-1/s}\} du d\lambda \\ &\leq 1 + C \int_1^\infty \lambda^{-t/s} d\lambda < \infty. \end{aligned}$$

- ▶  $\therefore E[I_s(\mu)] < \infty$  whenever  $0 < s < \underline{\text{ind}}U$ .



# Theorems of Pruitt and Taylor

- ▶  $I_s(\mu) = \int_0^1 \int_0^1 |X(u) - X(v)|^{-s} du dv.$
- ▶  $E[I_s(\mu)] = 2 \int_0^1 E(|X(u)|^{-s}) du.$
- ▶ If  $s < \underline{\text{ind}}U$ , then  $\int_0^1 P\{|X(u)| \leq \epsilon\} du = O(\epsilon^s)$ . Therefore, for all  $0 < s < t < \underline{\text{ind}}U$ ,

$$\begin{aligned} \int_0^1 E(|X(u)|^{-s}) du &= \int_0^1 \int_0^\infty P\{|X(u)|^{-s} > \lambda\} d\lambda du \\ &\leq 1 + \int_1^\infty \int_0^1 P\{|X(u)| \leq \lambda^{-1/s}\} du d\lambda \\ &\leq 1 + C \int_1^\infty \lambda^{-t/s} d\lambda < \infty. \end{aligned}$$

- ▶  $\therefore E[I_s(\mu)] < \infty$  whenever  $0 < s < \underline{\text{ind}}U$ .
- ▶  $\therefore \dim_H X([0, 1]) \geq \underline{\text{ind}}U$ . ■



# The Minkowski dimension

Need an analogue of Frostman's theorem.

## Theorem (Hu and Taylor)

Suppose  $F \subset \mathbf{R}^d$  is bounded measurable, and  $\exists$  probability measure  $\mu$  on  $F$  and  $s > 0$  such that

$$\liminf_{\epsilon \rightarrow 0} \frac{1}{\epsilon^s} \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy) < \infty.$$

Then,  $\overline{\dim}_M F \geq s$ .



# The Minkowski dimension

Need an analogue of Frostman's theorem.

## Theorem (Hu and Taylor)

Suppose  $F \subset \mathbf{R}^d$  is bounded measurable, and  $\exists$  probability measure  $\mu$  on  $F$  and  $s > 0$  such that

$$\liminf_{\epsilon \rightarrow 0} \frac{1}{\epsilon^s} \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy) < \infty.$$

Then,  $\overline{\dim}_M F \geq s$ .



# The Minkowski dimension

Need an analogue of Frostman's theorem.

## Theorem (Hu and Taylor)

Suppose  $F \subset \mathbf{R}^d$  is bounded measurable, and  $\exists$  probability measure  $\mu$  on  $F$  and  $s > 0$  such that

$$\liminf_{\epsilon \rightarrow 0} \frac{1}{\epsilon^s} \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy) < \infty.$$

Then,  $\overline{\dim}_M F \geq s$ .

**Proof:** If  $\mathcal{D}_n :=$  dyadic cubes, then (why?),

$$\iint \mathbf{1}\{|x - y| \leq 2^{-n} \sqrt{d}\} \mu(dx) \mu(dy)$$



# The Minkowski dimension

Need an analogue of Frostman's theorem.

## Theorem (Hu and Taylor)

Suppose  $F \subset \mathbf{R}^d$  is bounded measurable, and  $\exists$  probability measure  $\mu$  on  $F$  and  $s > 0$  such that

$$\liminf_{\epsilon \rightarrow 0} \frac{1}{\epsilon^s} \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy) < \infty.$$

Then,  $\overline{\dim}_M F \geq s$ .

**Proof:** If  $\mathcal{D}_n :=$  dyadic cubes, then (why?),

$$\begin{aligned} & \iint \mathbf{1}\{|x - y| \leq 2^{-n} \sqrt{d}\} \mu(dx) \mu(dy) \\ & \geq \sum_{I \in \mathcal{D}_n} |\mu(I)|^2 \mathbf{1}\{I \cap F \neq \emptyset\} \end{aligned}$$





# The Minkowski dimension

Need an analogue of Frostman's theorem.

## Theorem (Hu and Taylor)

Suppose  $F \subset \mathbf{R}^d$  is bounded measurable, and  $\exists$  probability measure  $\mu$  on  $F$  and  $s > 0$  such that

$$\liminf_{\epsilon \rightarrow 0} \frac{1}{\epsilon^s} \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy) < \infty.$$

Then,  $\overline{\dim}_M F \geq s$ .

**Proof:** If  $\mathcal{D}_n :=$  dyadic cubes, then (why?),

$$\begin{aligned} & \iint \mathbf{1}\{|x - y| \leq 2^{-n} \sqrt{d}\} \mu(dx) \mu(dy) \\ & \geq \sum_{I \in \mathcal{D}_n} |\mu(I)|^2 \mathbf{1}\{I \cap F \neq \emptyset\} \geq \frac{1}{N_n(F)} \end{aligned} \quad [\text{Cauchy-Schwarz}]$$



# The Minkowski dimension

Need an analogue of Frostman's theorem.

## Theorem (Hu and Taylor)

Suppose  $F \subset \mathbf{R}^d$  is bounded measurable, and  $\exists$  probability measure  $\mu$  on  $F$  and  $s > 0$  such that

$$\liminf_{\epsilon \rightarrow 0} \frac{1}{\epsilon^s} \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy) < \infty.$$

Then,  $\overline{\dim}_M F \geq s$ .

**Proof:** If  $\mathcal{D}_n :=$  dyadic cubes, then (why?),

$$\begin{aligned} & \iint \mathbf{1}\{|x - y| \leq 2^{-n} \sqrt{d}\} \mu(dx) \mu(dy) \\ & \geq \sum_{I \in \mathcal{D}_n} |\mu(I)|^2 \mathbf{1}\{I \cap F \neq \emptyset\} \geq \frac{1}{N_n(F)} \end{aligned} \quad \text{[Cauchy-Schwarz]}$$

Therefore,  $N_n(F) \geq c2^{ns}$  i.o. ■



## End of proof ( $\overline{\dim}_M$ )

Now we prove that  $\overline{\dim}_M X([0, 1]) \geq \overline{\text{ind}} U$  a.s., and finish the proof of the theorem.

$$\blacktriangleright \mu(A) := \int_0^1 \mathbf{1}_A(X(s)) ds.$$



## End of proof ( $\overline{\dim}_M$ )

Now we prove that  $\overline{\dim}_M X([0, 1]) \geq \overline{\text{ind}}U$  a.s., and finish the proof of the theorem.

- ▶  $\mu(A) := \int_0^1 \mathbf{1}_A(X(s)) ds.$
- ▶  $E \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy)$   
 $\leq 2 \int_0^1 P\{|X(u)| \leq \epsilon\} du$



## End of proof ( $\overline{\dim}_M$ )

Now we prove that  $\overline{\dim}_M X([0, 1]) \geq \overline{\text{ind}}U$  a.s., and finish the proof of the theorem.

- ▶  $\mu(A) := \int_0^1 \mathbf{1}_A(X(s)) ds.$
- ▶  $E \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy)$   
 $\leq 2 \int_0^1 P\{|X(u)| \leq \epsilon\} du$



## End of proof ( $\overline{\dim}_M$ )

Now we prove that  $\overline{\dim}_M X([0, 1]) \geq \overline{\text{ind}}U$  a.s., and finish the proof of the theorem.

- ▶  $\mu(A) := \int_0^1 \mathbf{1}_A(X(s)) ds.$
- ▶  $\mathbf{E} \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy)$   
 $\leq 2 \int_0^1 \mathbf{P}\{|X(u)| \leq \epsilon\} du = 2U(B(0, \epsilon))$



## End of proof ( $\overline{\dim}_M$ )

Now we prove that  $\overline{\dim}_M X([0, 1]) \geq \overline{\text{ind}}U$  a.s., and finish the proof of the theorem.

- ▶  $\mu(A) := \int_0^1 \mathbf{1}_A(X(s)) ds.$
- ▶  $\mathbf{E} \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy)$   
 $\leq 2 \int_0^1 \mathbf{P}\{|X(u)| \leq \epsilon\} du = 2U(B(0, \epsilon)) \leq \epsilon^{\overline{\text{ind}}U + o(1)}.$



## End of proof ( $\overline{\dim}_M$ )

Now we prove that  $\overline{\dim}_M X([0, 1]) \geq \overline{\text{ind}}U$  a.s., and finish the proof of the theorem.

- ▶  $\mu(A) := \int_0^1 \mathbf{1}_A(X(s)) ds.$
- ▶  $E \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy)$   
 $\leq 2 \int_0^1 P\{|X(u)| \leq \epsilon\} du = 2U(B(0, \epsilon)) \leq \epsilon^{\overline{\text{ind}}U + o(1)}.$
- ▶  $\therefore$  if  $0 < s < \overline{\text{ind}}U$ , Fatou's lemma  $\Rightarrow$

$$\liminf_{\epsilon \rightarrow 0} \frac{1}{\epsilon^s} \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy) = 0 \quad \text{a.s.}$$





## End of proof ( $\overline{\dim}_M$ )

Now we prove that  $\overline{\dim}_M X([0, 1]) \geq \overline{\text{ind}}U$  a.s., and finish the proof of the theorem.

- ▶  $\mu(A) := \int_0^1 \mathbf{1}_A(X(s)) ds.$
- ▶  $E \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy)$   
 $\leq 2 \int_0^1 P\{|X(u)| \leq \epsilon\} du = 2U(B(0, \epsilon)) \leq \epsilon^{\overline{\text{ind}}U + o(1)}.$
- ▶  $\therefore$  if  $0 < s < \overline{\text{ind}}U$ , Fatou's lemma  $\Rightarrow$

$$\liminf_{\epsilon \rightarrow 0} \frac{1}{\epsilon^s} \iint \mathbf{1}\{|x - y| \leq \epsilon\} \mu(dx) \mu(dy) = 0 \quad \text{a.s.}$$

- ▶ As  $\mu$  is a probab. meas. on  $X([0, 1])$ , the previous theorem implies that  $\overline{\dim}_M X([0, 1]) \geq \overline{\text{ind}}U$  a.s. ■



## Example: Brownian motion

- ▶  $X =$  Brownian motion on  $\mathbf{R}^d$ . Note that

$$U(B(0, \epsilon)) = \int_0^1 \mathbf{P}\{|X(s)| \leq \epsilon\} ds$$



## Example: Brownian motion

- ▶  $X =$  Brownian motion on  $\mathbf{R}^d$ . Note that

$$U(B(0, \epsilon)) = \int_0^1 \mathbf{P}\{|X(s)| \leq \epsilon\} ds$$



## Example: Brownian motion

- ▶  $X =$  Brownian motion on  $\mathbf{R}^d$ . Note that

$$\begin{aligned}U(B(0, \epsilon)) &= \int_0^1 \mathbf{P}\{|X(s)| \leq \epsilon\} ds \\ &= \int_0^1 \mathbf{P}\left\{|X(1)| \leq \frac{\epsilon}{\sqrt{s}}\right\} ds\end{aligned}$$



## Example: Brownian motion

- ▶  $X =$  Brownian motion on  $\mathbf{R}^d$ . Note that

$$\begin{aligned}U(B(0, \epsilon)) &= \int_0^1 \mathbf{P}\{|X(s)| \leq \epsilon\} ds \\&= \int_0^1 \mathbf{P}\left\{|X(1)| \leq \frac{\epsilon}{\sqrt{s}}\right\} ds \\&\asymp \int_0^1 \left(\frac{\epsilon}{\sqrt{s}} \wedge 1\right)^d ds.\end{aligned}$$



## Example: Brownian motion

- ▶  $X =$  Brownian motion on  $\mathbf{R}^d$ . Note that

$$\begin{aligned}U(B(0, \epsilon)) &= \int_0^1 \mathbf{P}\{|X(s)| \leq \epsilon\} ds \\ &= \int_0^1 \mathbf{P}\left\{|X(1)| \leq \frac{\epsilon}{\sqrt{s}}\right\} ds \\ &\asymp \int_0^1 \left(\frac{\epsilon}{\sqrt{s}} \wedge 1\right)^d ds.\end{aligned}$$

- ▶ If  $d = 1$ , then this is of sharp order  $\epsilon$ .



## Example: Brownian motion

- ▶  $X =$  Brownian motion on  $\mathbf{R}^d$ . Note that

$$\begin{aligned}U(B(0, \epsilon)) &= \int_0^1 \mathbf{P}\{|X(s)| \leq \epsilon\} ds \\&= \int_0^1 \mathbf{P}\left\{|X(1)| \leq \frac{\epsilon}{\sqrt{s}}\right\} ds \\&\asymp \int_0^1 \left(\frac{\epsilon}{\sqrt{s}} \wedge 1\right)^d ds.\end{aligned}$$

- ▶ If  $d = 1$ , then this is of sharp order  $\epsilon$ .
- ▶ If  $d = 2$ , then this is of sharp order  $\epsilon^2 \log(1/\epsilon)$ .



## Example: Brownian motion

- ▶  $X =$  Brownian motion on  $\mathbf{R}^d$ . Note that

$$\begin{aligned}U(B(0, \epsilon)) &= \int_0^1 \mathbf{P}\{|X(s)| \leq \epsilon\} ds \\&= \int_0^1 \mathbf{P}\left\{|X(1)| \leq \frac{\epsilon}{\sqrt{s}}\right\} ds \\&\asymp \int_0^1 \left(\frac{\epsilon}{\sqrt{s}} \wedge 1\right)^d ds.\end{aligned}$$

- ▶ If  $d = 1$ , then this is of sharp order  $\epsilon$ .
- ▶ If  $d = 2$ , then this is of sharp order  $\epsilon^2 \log(1/\epsilon)$ .
- ▶ If  $d \geq 3$ , then this is of sharp order  $\epsilon^2$ .





## Example: Brownian motion

- ▶  $X =$  Brownian motion on  $\mathbf{R}^d$ . Note that

$$\begin{aligned}U(B(0, \epsilon)) &= \int_0^1 \mathbf{P}\{|X(s)| \leq \epsilon\} ds \\ &= \int_0^1 \mathbf{P}\left\{|X(1)| \leq \frac{\epsilon}{\sqrt{s}}\right\} ds \\ &\asymp \int_0^1 \left(\frac{\epsilon}{\sqrt{s}} \wedge 1\right)^d ds.\end{aligned}$$

- ▶ If  $d = 1$ , then this is of sharp order  $\epsilon$ .
- ▶ If  $d = 2$ , then this is of sharp order  $\epsilon^2 \log(1/\epsilon)$ .
- ▶ If  $d \geq 3$ , then this is of sharp order  $\epsilon^2$ .
- ▶  $\therefore \overline{\text{ind}}U = \underline{\text{ind}}U = \min(d, 2)$ .



# Problems

1. Suppose  $\mu$  is a Borel measure on  $\mathbf{R}^d$ , and  $\sup_{I \in \mathcal{D}_n} \mu(I) = O(2^{-ns})$  for some  $s > 0$ . Then, prove that  $\sup_{x \in \mathbf{R}^d} \mu(B(x, \epsilon)) = O(\epsilon^s)$ .



# Problems

1. Suppose  $\mu$  is a Borel measure on  $\mathbf{R}^d$ , and  $\sup_{I \in \mathcal{D}_n} \mu(I) = O(2^{-ns})$  for some  $s > 0$ . Then, prove that  $\sup_{x \in \mathbf{R}^d} \mu(B(x, \epsilon)) = O(\epsilon^s)$ .
2. Let  $F$  be a Borel subset of  $\mathbf{R}^d$ , and define the *lower Minkowski dimension* of  $F$  to be

$$\underline{\dim}_M F := \liminf_{n \rightarrow \infty} \frac{\log_2 N_n(F)}{n},$$

where  $N_n(F)$  denotes the number of dyadic cubes that intersect  $F$ .  
Prove:



# Problems

1. Suppose  $\mu$  is a Borel measure on  $\mathbf{R}^d$ , and  $\sup_{I \in \mathcal{D}_n} \mu(I) = O(2^{-ns})$  for some  $s > 0$ . Then, prove that  $\sup_{x \in \mathbf{R}^d} \mu(B(x, \epsilon)) = O(\epsilon^s)$ .
2. Let  $F$  be a Borel subset of  $\mathbf{R}^d$ , and define the *lower Minkowski dimension* of  $F$  to be

$$\underline{\dim}_M F := \liminf_{n \rightarrow \infty} \frac{\log_2 N_n(F)}{n},$$

where  $N_n(F)$  denotes the number of dyadic cubes that intersect  $F$ .  
Prove:

$$2.1 \quad \dim_H F \leq \underline{\dim}_M F \leq \overline{\dim}_M F.$$



# Problems

1. Suppose  $\mu$  is a Borel measure on  $\mathbf{R}^d$ , and  $\sup_{I \in \mathcal{D}_n} \mu(I) = O(2^{-ns})$  for some  $s > 0$ . Then, prove that  $\sup_{x \in \mathbf{R}^d} \mu(B(x, \epsilon)) = O(\epsilon^s)$ .
2. Let  $F$  be a Borel subset of  $\mathbf{R}^d$ , and define the *lower Minkowski dimension* of  $F$  to be

$$\underline{\dim}_M F := \liminf_{n \rightarrow \infty} \frac{\log_2 N_n(F)}{n},$$

where  $N_n(F)$  denotes the number of dyadic cubes that intersect  $F$ .  
Prove:

$$2.1 \quad \dim_H F \leq \underline{\dim}_M F \leq \overline{\dim}_M F.$$

$$2.2 \quad \underline{\dim}_M X([0, 1]) = \dim_H X([0, 1]) \quad \forall \text{ Lévy processes } X.$$



## More advanced problems

1. Prove that

$$P\{X(s) \in B(a, \epsilon) \text{ for some } s \leq \zeta\} \geq \frac{U(B(a, \epsilon))}{U(B(0, 2\epsilon))}.$$



## More advanced problems

1. Prove that

$$P\{X(s) \in B(a, \epsilon) \text{ for some } s \leq \zeta\} \geq \frac{U(B(a, \epsilon))}{U(B(0, 2\epsilon))}.$$

2. Prove that

$$\dim_H X([0, 1]) = \sup \left\{ s > 0 : \int_0^1 E(|X(u)|^{-s}) du < \infty \right\}.$$



## More advanced problems

1. Prove that

$$P\{X(s) \in B(a, \epsilon) \text{ for some } s \leq \zeta\} \geq \frac{U(B(a, \epsilon))}{U(B(0, 2\epsilon))}.$$

2. Prove that

$$\dim_{\text{H}} X([0, 1]) = \sup \left\{ s > 0 : \int_0^1 E(|X(u)|^{-s}) du < \infty \right\}.$$

3. Let  $X :=$  isotropic stable, index  $\alpha \in (0, 2]$ :

$E \exp(i\xi \cdot X(t)) = \exp(-t\|\xi\|^\alpha)$ . Prove:

$\dim_{\text{H}} X([0, 1]) = \overline{\dim}_{\text{M}} X([0, 1]) = \min(d, \alpha)$  a.s.





## More advanced problems

1. Prove that

$$P\{X(s) \in B(a, \epsilon) \text{ for some } s \leq \zeta\} \geq \frac{U(B(a, \epsilon))}{U(B(0, 2\epsilon))}.$$

2. Prove that

$$\dim_{\text{H}} X([0, 1]) = \sup \left\{ s > 0 : \int_0^1 E(|X(u)|^{-s}) du < \infty \right\}.$$

3. Let  $X$  := isotropic stable, index  $\alpha \in (0, 2]$ :

$E \exp(i\xi \cdot X(t)) = \exp(-t\|\xi\|^\alpha)$ . Prove:

$\dim_{\text{H}} X([0, 1]) = \overline{\dim}_{\text{M}} X([0, 1]) = \min(d, \alpha)$  a.s.

4. (Hard) Prove that:



## More advanced problems

1. Prove that

$$P\{X(s) \in B(a, \epsilon) \text{ for some } s \leq \zeta\} \geq \frac{U(B(a, \epsilon))}{U(B(0, 2\epsilon))}.$$

2. Prove that

$$\dim_{\text{H}} X([0, 1]) = \sup \left\{ s > 0 : \int_0^1 E(|X(u)|^{-s}) du < \infty \right\}.$$

3. Let  $X$  := isotropic stable, index  $\alpha \in (0, 2]$ :

$E \exp(i\xi \cdot X(t)) = \exp(-t\|\xi\|^\alpha)$ . Prove:

$\dim_{\text{H}} X([0, 1]) = \overline{\dim}_{\text{M}} X([0, 1]) = \min(d, \alpha)$  a.s.

4. (Hard) Prove that:

4.1  $U(B(a, \epsilon)) \leq U(B(0, 2\epsilon))$ .



# More advanced problems

1. Prove that

$$P\{X(s) \in B(a, \epsilon) \text{ for some } s \leq \zeta\} \geq \frac{U(B(a, \epsilon))}{U(B(0, 2\epsilon))}.$$

2. Prove that

$$\dim_{\text{H}} X([0, 1]) = \sup \left\{ s > 0 : \int_0^1 E(|X(u)|^{-s}) du < \infty \right\}.$$

3. Let  $X$  := isotropic stable, index  $\alpha \in (0, 2]$ :

$E \exp(i\xi \cdot X(t)) = \exp(-t\|\xi\|^\alpha)$ . Prove:

$\dim_{\text{H}} X([0, 1]) = \overline{\dim}_{\text{M}} X([0, 1]) = \min(d, \alpha)$  a.s.

4. (Hard) Prove that:

4.1  $U(B(a, \epsilon)) \leq U(B(0, 2\epsilon))$ .

4.2  $U(B(0, 2\epsilon)) \leq 16^d U(B(0, \epsilon))$ .

[Volume doubling]

