

Random walks in cones: critical exponents

Lecture #2

Analytic and probabilistic tools for lattice path enumeration

KILIAN RASCHEL



77th Séminaire Lotharingien de Combinatoire
September 13, 2016
Strobl, Austria

Introduction

Dimension 1: examples & limits

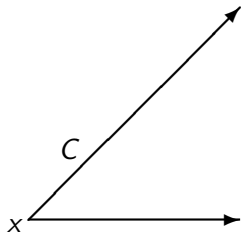
Central idea in dimension ≥ 2 : approximation by Brownian motion

Application #1: excursions

Application #2: walks with prescribed length

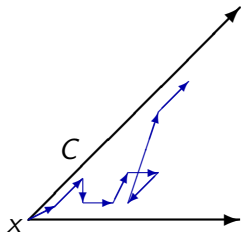
Random processes (RW & BM) in cones

First exit time from a cone C



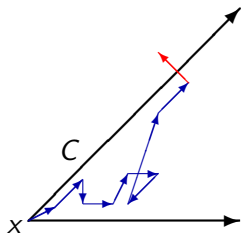
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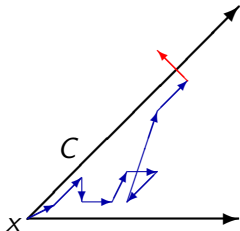
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- ▷ $\tau_C = \inf\{n \in \mathbf{N} : S(n) \notin C\}$ (S RW)
- ▷ $T_C = \inf\{t \in \mathbf{R}_+ : B(t) \notin C\}$ (B BM)

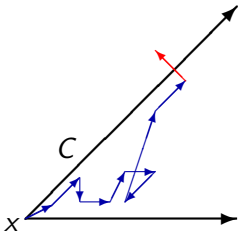


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Persistence probabilities \leadsto total number of walks

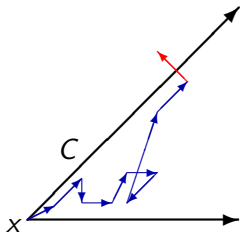
▷ $\mathbf{P}_x[\tau_C > n] \sim \kappa \cdot V(x) \cdot \rho^n \cdot n^{-\alpha}$

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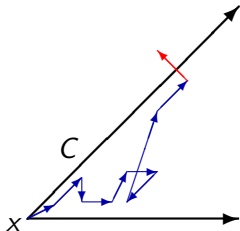
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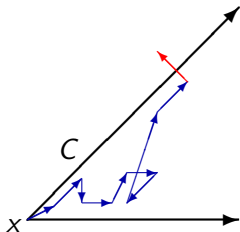
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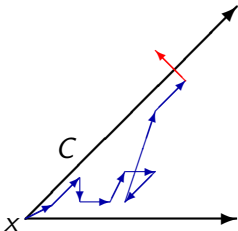
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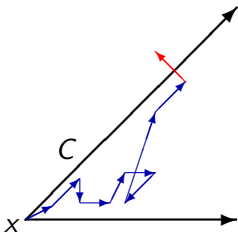
Local limit theorems \rightsquigarrow excursions

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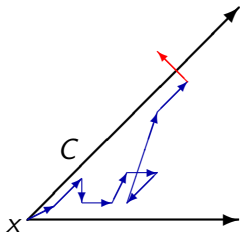
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Aim of the talk: understanding the critical exponents α

Definition of random walks & motivations

Random walk on \mathbb{Z}^d

▷ A *random walk* $\{S(n)\}_{n \geq 0}$ is

$$S(n) = x + X(1) + \cdots + X(n),$$

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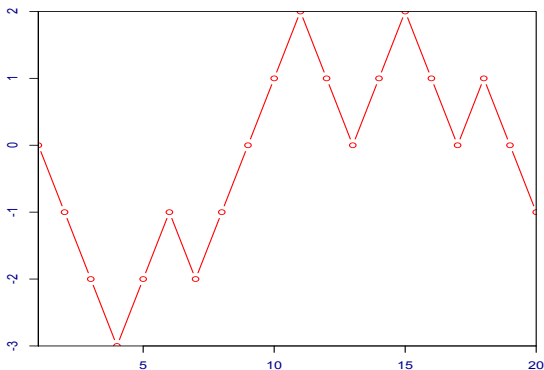
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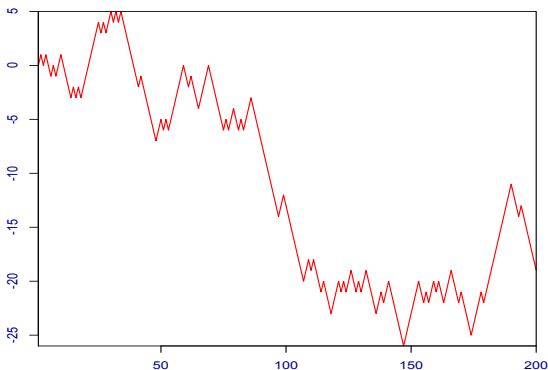
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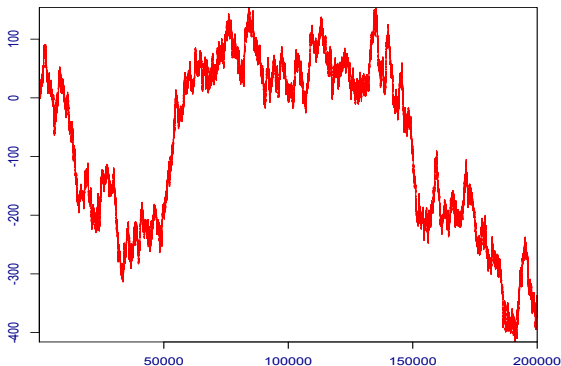
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Motivations

- ▷ *Persistence probabilities* in statistical physics
- ▷ Constructing *processes conditioned* never to leave cones

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Motivations

- ▷ *Persistence probabilities* in statistical physics
- ▷ Constructing *processes conditioned* never to leave cones
- ▷ *Asymptotics* of numbers of walks
- ▷ *Transcendental nature* of functions counting walks in cones
↪ Alin Bostan's course at AEC
- ▷ Important & combinatorial cones (quarter/half/slit plane, orthants, Weyl chambers, etc.)

Introduction

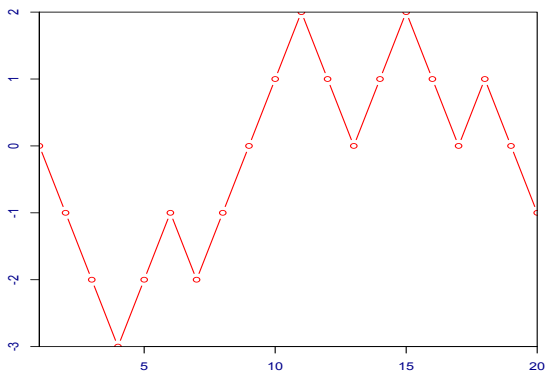
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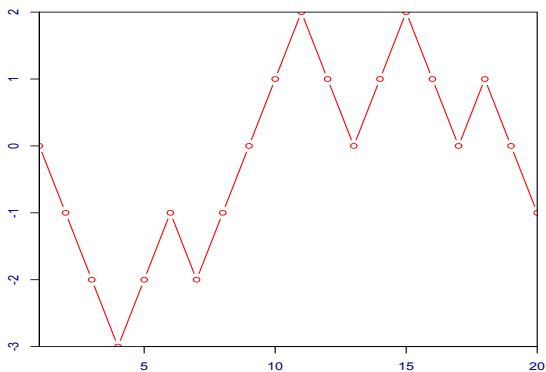
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Non-constrained walk with $\mathcal{S} = \{-1, 1\}$



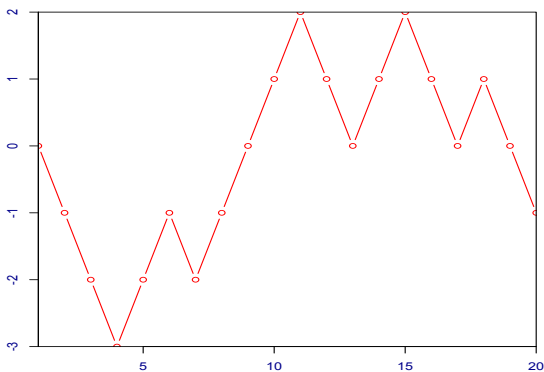
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▷ $\#\{x \xrightarrow{n} \mathbf{Z}\} = 2^n$

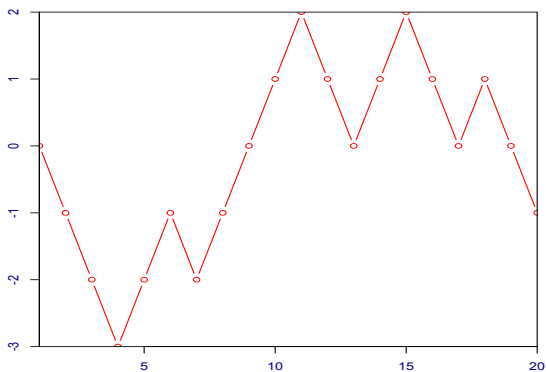
Walk \rightsquigarrow Exponent 0

Non-constrained walk with $\mathfrak{S} = \{-1, 1\}$



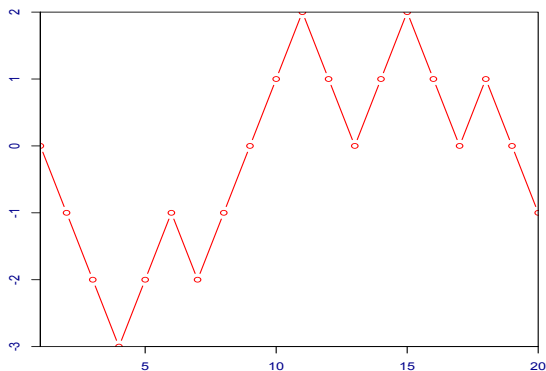
- ▷ $\#\{x \xrightarrow{n} \mathbf{z}\} = 2^n$ Walk \rightsquigarrow Exponent 0
- ▷ $\#\{x \xrightarrow{n} y\} = \binom{n}{\frac{n+(y-x)}{2}} \sim \sqrt{\frac{2}{\pi}} \frac{2^n}{\sqrt{n}}$ Bridge \rightsquigarrow Exponent $\frac{1}{2}$

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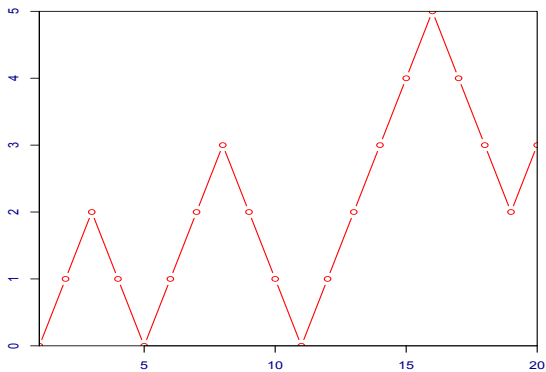
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- ▷ $\sum \frac{1}{\sqrt{n}} = \infty$: *recurrence* of the simple random walk in \mathbf{Z}

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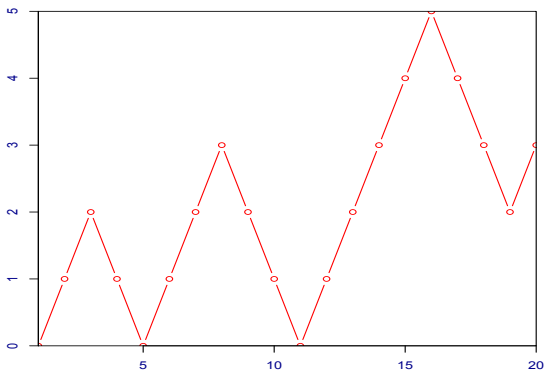


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- ▷ $\sum \frac{1}{\sqrt{n}} = \infty$: recurrence of the simple random walk in \mathbf{Z}
- ▷ Constant $\sqrt{\frac{2}{\pi}}$ independent of x & y in the asymptotics

Constrained walk with $\mathcal{S} = \{-1, 1\}$ (Dyck paths)



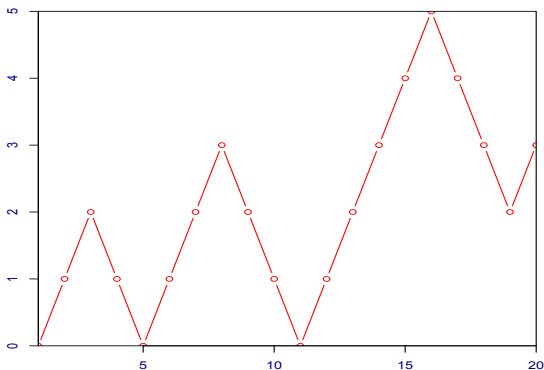
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▷ $\#_{\mathbf{N}}\{x \xrightarrow{n} \mathbf{N}\} \sim \frac{2^n}{\sqrt{n}}$

Meanders \rightsquigarrow Exponent $\frac{1}{2}$

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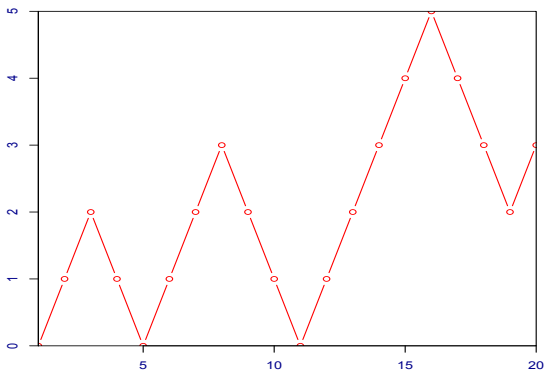


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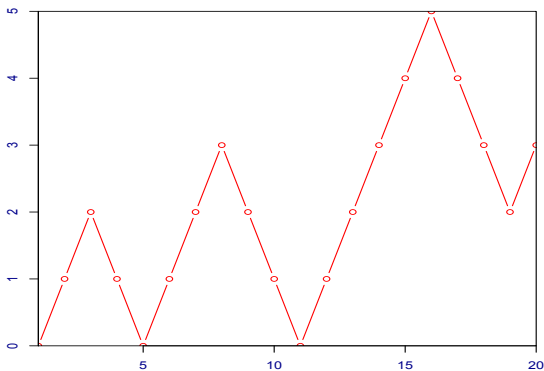
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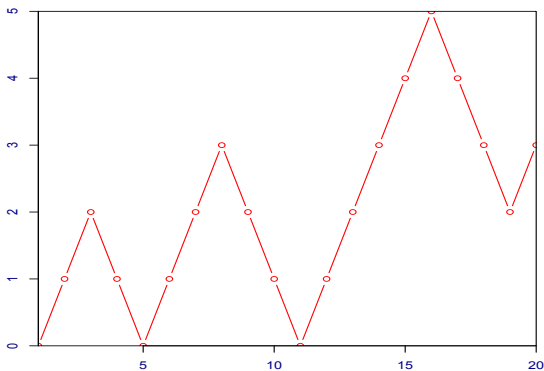
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▷ *Wiener-Hopf* techniques in probability theory

▷ See  Bousquet-Mélou & Petkovšek '00; Banderier & Flajolet '02

Beyond the algebraic exponents 0 , $\frac{1}{2}$ & $\frac{3}{2}$

Weighted models in dimension 1

Drift $\sum_{s \in \mathfrak{G}} s$ governs the exponents, which are still 0 , $\frac{1}{2}$ & $\frac{3}{2}$

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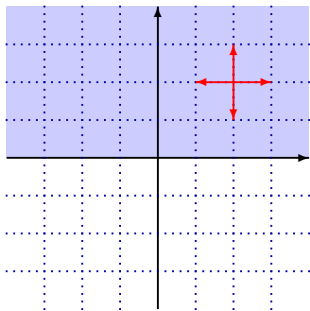
The simple walk in two-dimensional wedges

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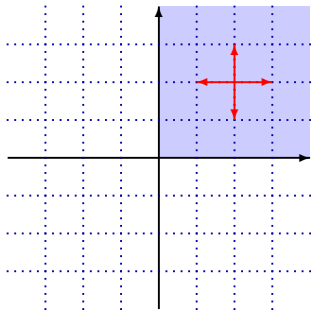
- ▷ **Half-plane:**
one-dimensional case
- ▷ Dyck paths
- ▷ Total number of walks:
↪ Exponent $\frac{1}{2}$
- ▷ Excursions:
↪ Exponent $2 = \frac{3}{2} + \frac{1}{2}$

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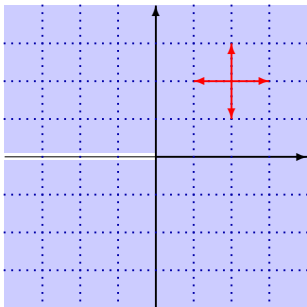
- ▷ **Quarter plane:** product of two one-dimensional cases
- ▷ Reflection principle
- ▷ Total number of walks:
 \rightsquigarrow **Exponent 1** $= \frac{1}{2} + \frac{1}{2}$
- ▷ Excursions:
 \rightsquigarrow **Exponent 3**

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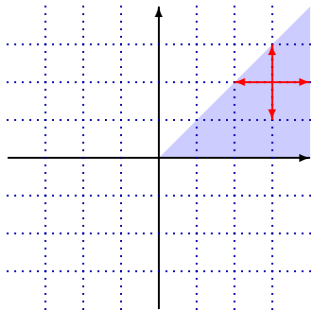
- ▷ **Slit plane:**
 - 📎 Bousquet-Mélou & Schaeffer '00
- ▷ Highly non-convex cone
- ▷ Total number of walks:
 - ↪ **Exponent $\frac{1}{4}$**
- ▷ Excursions:
 - ↪ **Exponent $\frac{3}{2}$**



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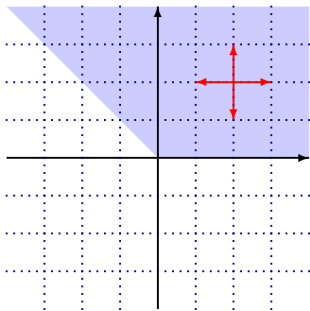
- ▷ 45° :  Gouyou-Beauchamps '86
- ▷ See  Bousquet-Mélou & Mishna '10
- ▷ Total number of walks:
 \rightsquigarrow Exponent 2
- ▷ Excursions:
 \rightsquigarrow Exponent 5


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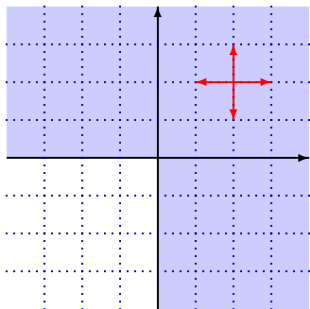
- ▷ 135° : Gessel
- ▷ See  Kauers, Koutschan & Zeilberger '09, etc.
- ▷ Total number of walks:
 - ↪ Exponent $\frac{2}{3}$
- ▷ Excursions:
 - ↪ Exponent $\frac{7}{3}$


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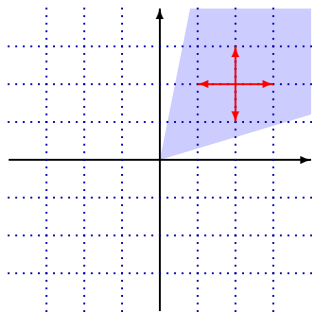
- ▷ Walks **avoiding a quadrant**
- ▷ See  Bousquet-Mélou '15;
Mustapha '15
- ▷ Total number of walks:
 \rightsquigarrow Exponent $\frac{1}{3}$
- ▷ Excursions:
 \rightsquigarrow Exponent $\frac{5}{3}$


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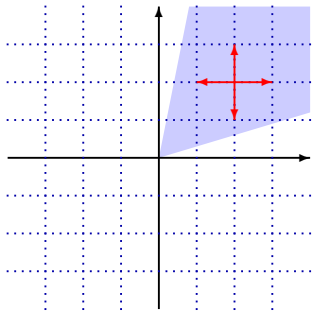
- ▷ Arbitrary **angular sector** θ
- ▷ See  Varopoulos '99; Denisov & Wachtel '15

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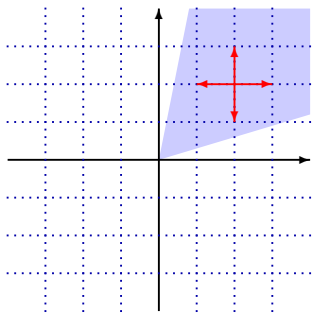
- ▷ Arbitrary **angular sector** θ
- ▷ See 📖 Varopoulos '99; Denisov & Wachtel '15
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Beyond the algebraic exponents 0, $\frac{1}{2}$ & $\frac{3}{2}$

Weighted models in dimension 1

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Conclusion: 1D case not enough

- ▷ Dramatic change of behavior: every exponent is possible!
- ▷ *Non-D-finite* behaviors (first observed by Varopoulos '99)

Introduction

Dimension 1: examples & limits

Central idea in dimension ≥ 2 : approximation by Brownian motion

Application #1: excursions

Application #2: walks with prescribed length

Brownian motion on R

Law of large numbers

$$\frac{X(1) + \cdots + X(n)}{n^1} \xrightarrow{\text{a.s.}} \mathbf{E}[X(1)]$$

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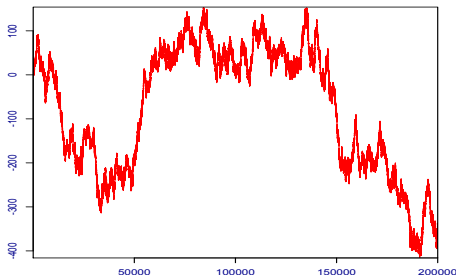
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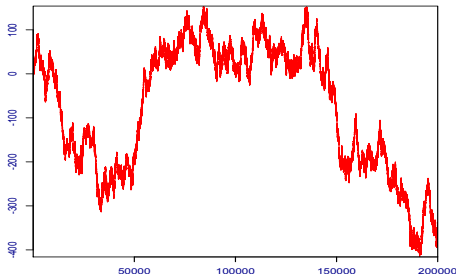
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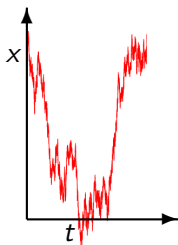
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Remainder of this section: computing $\alpha\{\text{BM}\}$ (easier)

Two derivations of the BM persistence probability in R

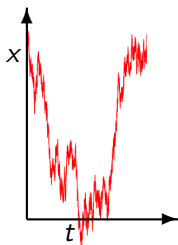
Reflection principle



$$\begin{aligned}\mathbf{P}_x[T_{(0,\infty)} > t] &= \mathbf{P}_0[\min_{0 \leq u \leq t} B(u) > -x] \\ &= \mathbf{P}_0[|B(t)| < x] \\ &= \frac{2}{\sqrt{2\pi t}} \int_0^x e^{-\frac{y^2}{2t}} dy\end{aligned}$$

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Heat equation

Function $g(t; x) = \mathbf{P}_x[T_{(0,\infty)} > t]$ satisfies

$$\left\{ \begin{array}{l} (\frac{\partial}{\partial t} - \frac{1}{2}\Delta) g(t; x) = 0, \quad \forall x \in (0, \infty), \quad \forall t \in (0, \infty) \\ g(0; x) = 1, \quad \forall x \in (0, \infty) \\ g(t; 0) = 0, \quad \forall t \in (0, \infty) \end{array} \right.$$

Dimension d : explicit expression for $\mathbf{P}_x[T_C > t]$

Heat equation

 Doob '55

For essentially *any domain* C in *any dimension* d , $\mathbf{P}_x[T_C > t]$ & $p^C(t; x, y)$ ($\mathbf{P}_x[T_C > t] = \int_C p^C(t; x, y) dy$) satisfy *heat equations*

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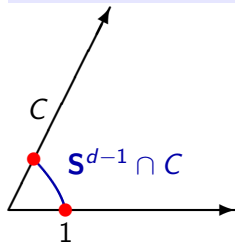
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$$\begin{cases} \Delta_{\mathbf{S}^{d-1}} m = -\lambda m & \text{in } \mathbf{S}^{d-1} \cap C \\ m = 0 & \text{in } \partial(\mathbf{S}^{d-1} \cap C) \end{cases}$$

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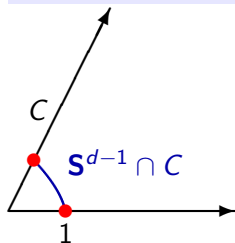
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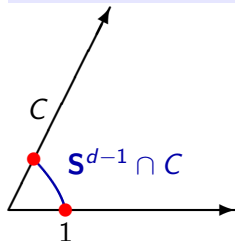
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$$\mathbf{P}_x[T_C > t] = \sum_{j=1}^{\infty} B_j(|x|^2/t) m_j(x/|x|)$$

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Exercise

Recover the **exponent** $\frac{\pi}{2\theta}$ of the persistence probability for a simple random walk in a two-dimensional wedge of opening angle θ

Introduction

Dimension 1: examples & limits

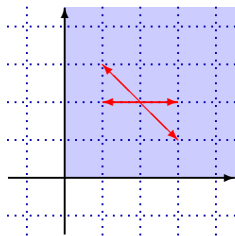
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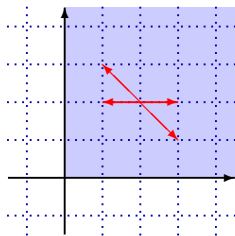
Example #1: Gouyou-Beauchamps model

In the quarter plane



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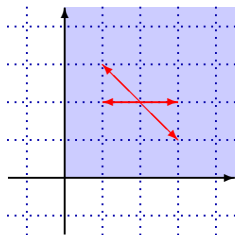


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$$\begin{aligned}\mathbf{E}[\mathbf{GB}] &= (1, 0) + (1, -1) + (-1, 0) + (-1, 1) \\ &= (0, 0)\end{aligned}$$

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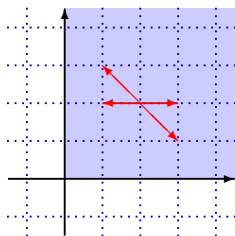
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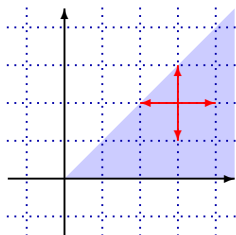


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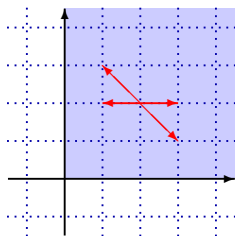
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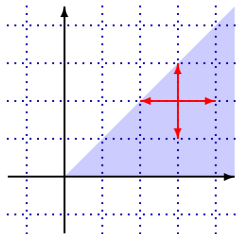


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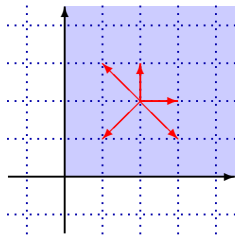
Changing the cone



- ▷ Wedge of angle $\theta = \frac{\pi}{4}$
- ▷ Total number of walks:
 \rightsquigarrow Exponent $\frac{\pi}{2\theta} = 2$
- ▷ Excursions:
 \rightsquigarrow Exponent $\frac{\pi}{\theta} + 1 = 5$

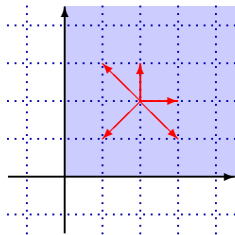
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A scarecrow



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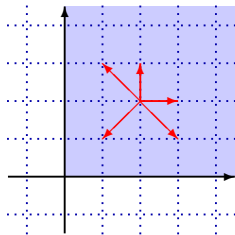
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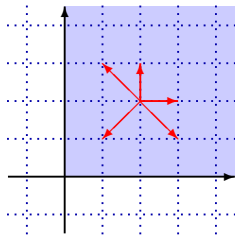
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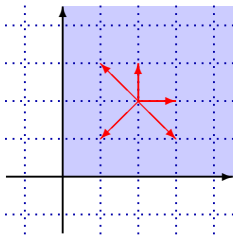
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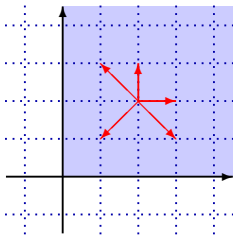
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 Bostan, R. & Salvy '14

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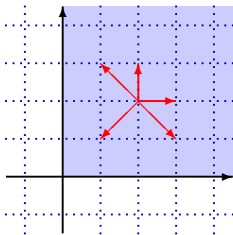
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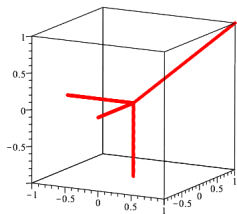
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- ▷ If $\sum_{s \in \mathcal{G}} s \neq 0$, first perform a *Cramér transform*

Three-dimensional models

Example: Kreweras 3D

Model with jumps:

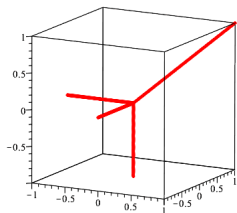


Three-dimensional models

Example: Kreweras 3D

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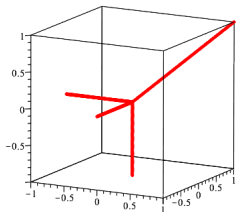
$$\text{Exponent } \alpha = 2\sqrt{\lambda_1 + \frac{1}{4}} - \frac{1}{2}$$



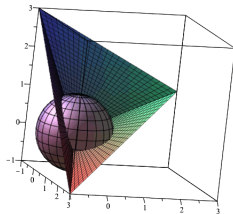
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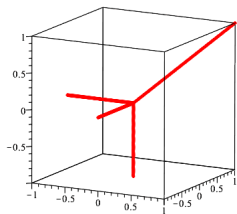
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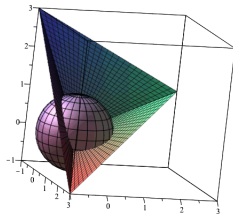
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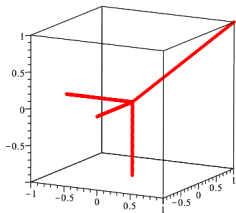


Value of λ_1 ? $\lambda_1 \in \mathbf{Q}$?

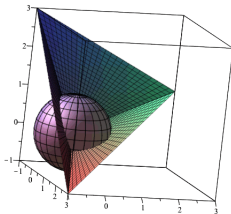
Three-dimensional models

Example: Kreweras 3D

Model with jumps:



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Value of λ_1 ? $\lambda_1 \in \mathbf{Q}$?

General theory (still to be done!)

▷ Classification & resolution of some finite group models

📎 Bostan, Bousquet-Mélou, Kauers & Melczer '16

▷ Asymptotic simulation

📎 Bacher, Kauers & Yatchak '16

↪ Conjectured Kreweras exponent: 3.3257569

▷ Equivalence finite group iff D-finite generating functions?

Introduction

Dimension 1: examples & limits

Central idea in dimension ≥ 2 : approximation by Brownian motion

Application #1: excursions

Application #2: walks with prescribed length

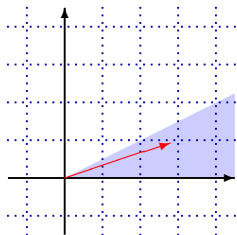
Non-universal exponents: six cases

Excursions: formula for α independent of the drift $\sum_{s \in \mathcal{G}} s$

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Excursions: formula for α independent of the drift $\sum_{s \in \mathcal{G}} s$

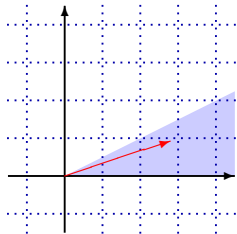
Case #1: interior drift



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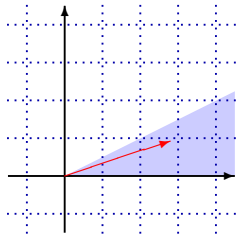


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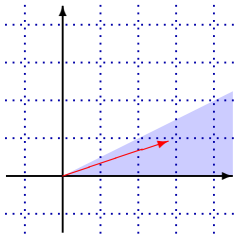


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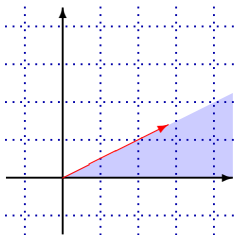
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Case #2: boundary drift

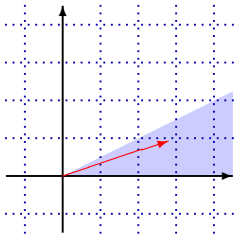


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- ▷ Exponent $\alpha = \frac{1}{2}$

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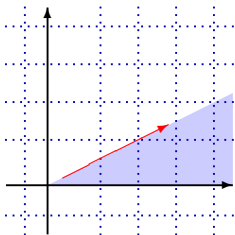
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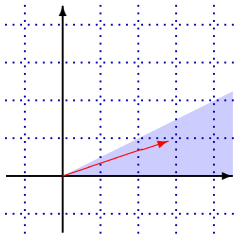


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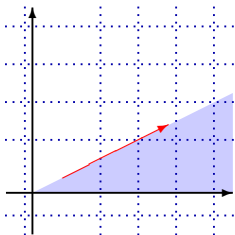
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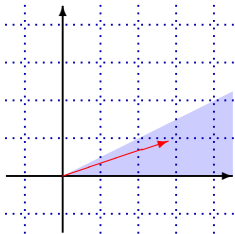


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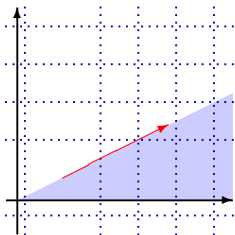
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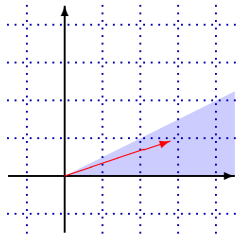


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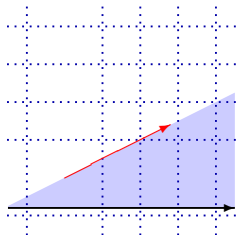
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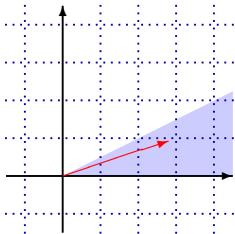


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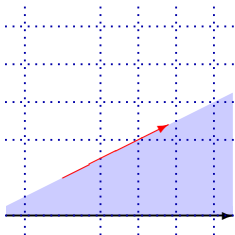
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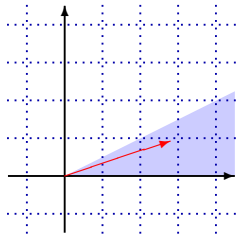


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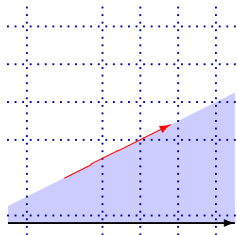
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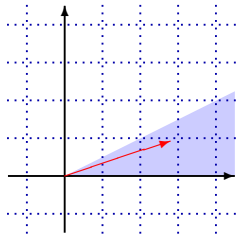


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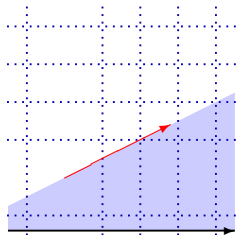
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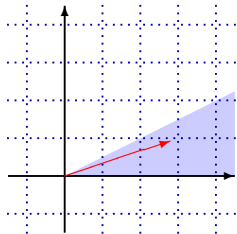


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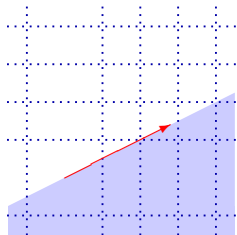
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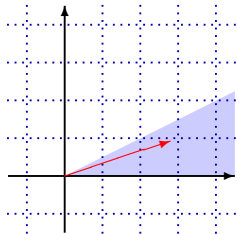


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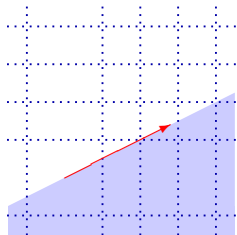
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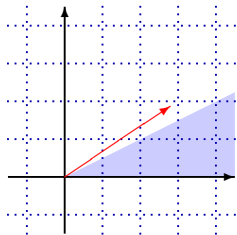
Case #2: boundary drift



- ▷ Half-plane case
- ▷ Exponent $\alpha = \frac{1}{2}$
- ▷ *Cannot* be used as a filter to detect non-D-finiteness
- ▷ Exponent $\alpha = \frac{i}{2}$ for non-smooth boundary

Non-universal exponents: six cases

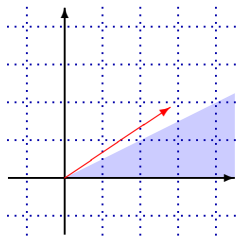
Case #3: directed drift



- ▷ Half-plane case
- ▷ Exponent $\alpha = \frac{3}{2}$
- ▷ *Cannot* be used as a filter to detect non-D-finiteness

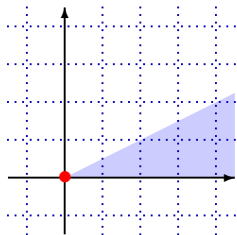
Non-universal exponents: six cases


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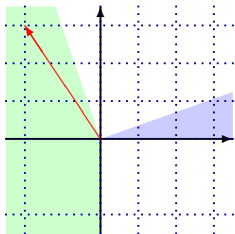
Case #4: zero drift




- ▷ See  Varopoulos '99; Denisov & Wachtel '15
- ▷ Exponent
$$\alpha_1 = 2\sqrt{\lambda_1 + \left(\frac{d}{2} - 1\right)^2} - \left(\frac{d}{2} - 1\right)$$
- ▷ *Can* be used as a filter to detect non-D-finiteness

Non-universal exponents: six cases

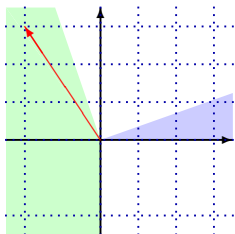
Case #5: polar interior drift




- ▷ See  Duraj '14
- ▷ Exponent $2\alpha_1 + 1$
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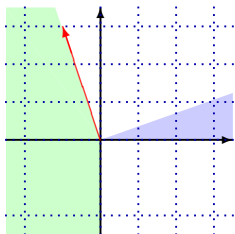
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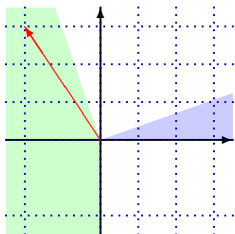
Case #6: polar boundary drift




- ▷ Exponent $\alpha_1 + 1$
- ▷ *Can* be used as a filter to detect non-D-finiteness

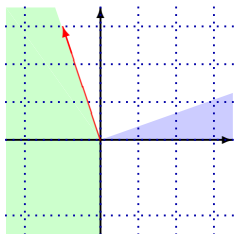
Non-universal exponents: six cases

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- ▷ Exponent $\alpha_1 + 1$
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Six-exponents-result: joint with R. Garbit & S. Mustapha

