

Ornstein-Zernike asymptotics in Statistical Mechanics

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based on joint works with M. Campanino and D. Ioffe

Contents

- 1 Effective random walk representation
- 2 Ornstein-Zernike theory
- 3 Some mathematical results
- 4 Overview of the approach

Random walks in Physics

Random walks are often used in Physics as simple effective models for more complicated objects.

Three examples to be discussed:

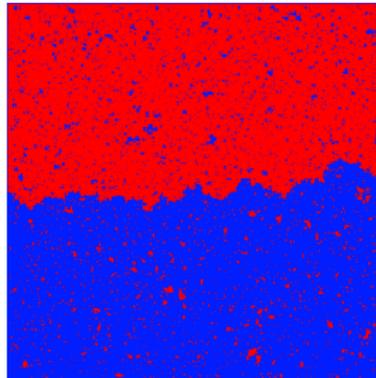
- Interfaces in 2d
- Subcritical clusters
- Stretched polymers

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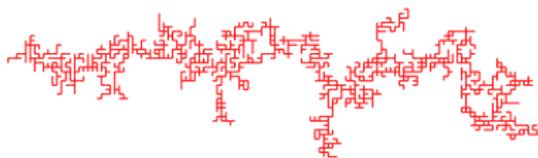


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- **Subcritical clusters**
- Stretched polymers

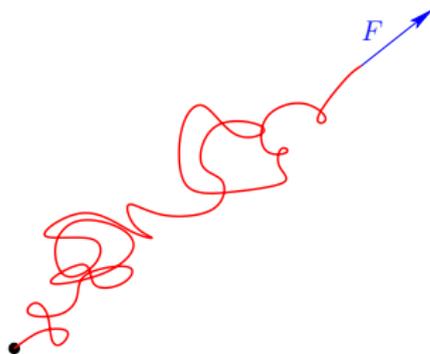


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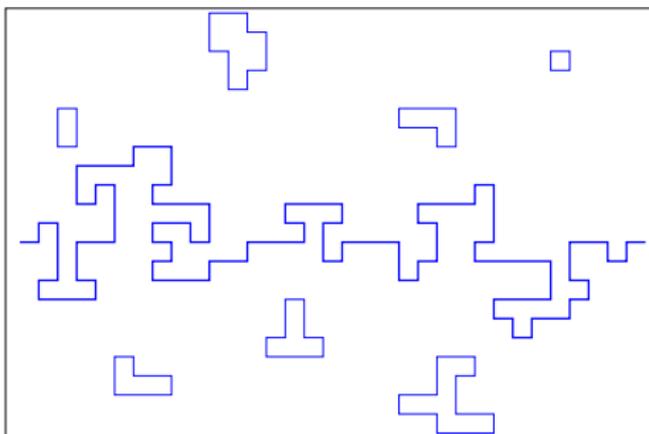


Example: 1D Interface

Problem: Analysis of the statistical properties of an interface separating two equilibrium phases.

Complicated geometry!

Is it possible to analyse instead a simpler model?

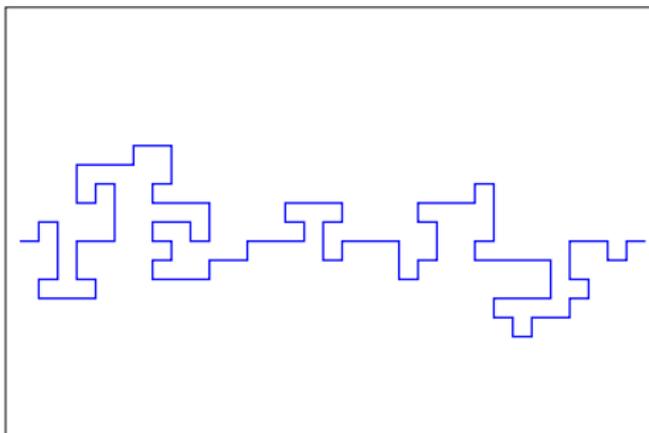


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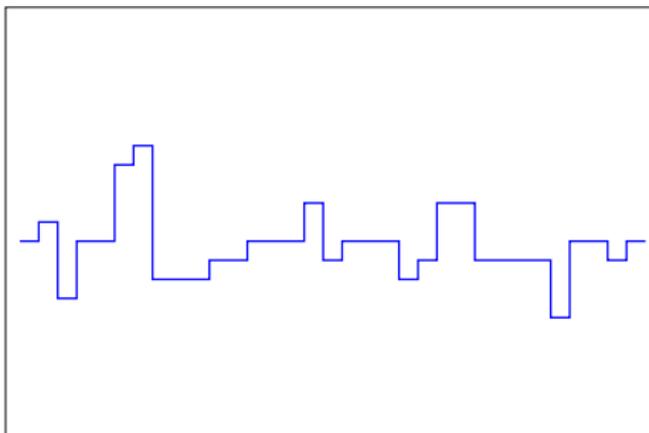


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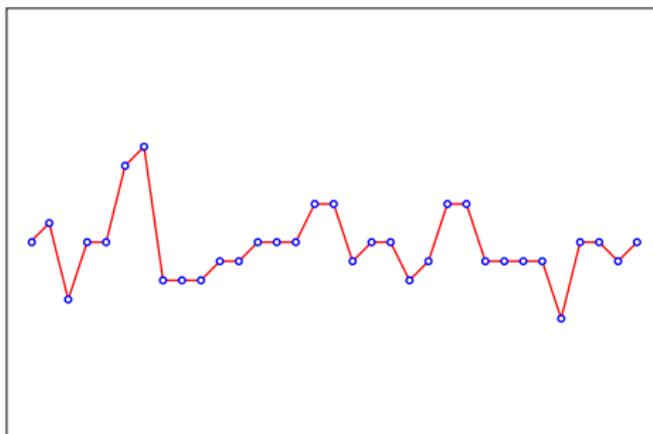


Example: 1D Interface

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Heuristic arguments

- Structure should be simple at microscales large compared to the correlation length.
- Large scale fluctuations should display universal (Brownian) asymptotics.

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Ornstein-Zernike theory

Proposed in 1914 by L. S. Ornstein and F. Zernike.



Aim: Determine the (large distance) behaviour of density-density correlations in simple fluid, away from criticality.

Ornstein-Zernike theory

OZ equation:

$$G(x_1 - x_2) = C(x_1 - x_2) + \rho \int C(x_1 - x_3)G(x_3 - x_2) dx_3$$

- $g(x)$: pair correlation function
- $G(x) = g(x) - 1$: net correlation function
- $C(x)$: direct correlation function

Ornstein-Zernike theory

Fourier transform:

$$\widehat{G}(k) = \frac{\widehat{C}(k)}{1 - \rho \widehat{C}(k)}$$

Ornstein-Zernike theory

Fourier transform:

$$\widehat{G}(k) = \frac{\widehat{C}(k)}{1 - \rho \widehat{C}(k)}$$

Crucial assumption: Separation of masses

C has smaller range than G (faster exp. decay)

Ornstein-Zernike theory

Fourier transform:

$$\widehat{G}(k) = \frac{\widehat{C}(k)}{1 - \rho \widehat{C}(k)}$$

Possible to expand:

$$\widehat{C}(k) \approx \widehat{C}(0) - R^2 |k|^2$$

Ornstein-Zernike theory

Fourier transform:

$$\hat{G}(k) \approx \frac{\hat{C}(0)}{\rho R^2(\kappa^2 + |k|^2)}$$

Possible to expand:

$$\hat{C}(k) \approx \hat{C}(0) - R^2|k|^2$$

Ornstein-Zernike theory

This yields

OZ asymptotics

$$G(x) \approx \frac{A}{|x|^{(d-1)/2}} e^{-\kappa|x|}$$

(valid as $|x| \rightarrow \infty$ with κ fixed)

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Goal: Rigorous understanding of both (closely related) problems from first principles. Already well understood 30 years ago, in **perturbative** regime:

- **OZ:** [Abraham, Kunz '77], [Paes-Leme '78], [Bricmont, Fröhlich '85], etc.
- **Scaling of 2D Ising interface:** [Gallavotti '72], [Higuchi '79], [Bricmont, Fröhlich, Pfister '81], etc.

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Nonperturbative results for very simple models: SAW [Chayes, Chayes '86, Ioffe '98], percolation [Campanino, Chayes, Chayes '91]. Very model-specific approaches.

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Robust nonperturbative approach: percolation [Campanino, Ioffe '02], Ising model [Campanino, Ioffe, V '03], random-cluster model [Campanino, Ioffe, V '08], selfinteracting polymers [Ioffe, V '08].

EXAMPLE #1

Interfaces in 2D systems

q -states Potts model

- $\Lambda \in \mathbb{Z}^2$
- $\sigma_i \in \{1, \dots, q\}, \forall i \in \Lambda$
- $\beta \geq 0$

$$\pi_{\Lambda}^{\beta, q}(\sigma) \propto \exp\left(\beta \sum_{i \sim j} \delta_{\sigma_i, \sigma_j}\right)$$

In particular, for $q = 2$: **Ising model**

q -states Potts model

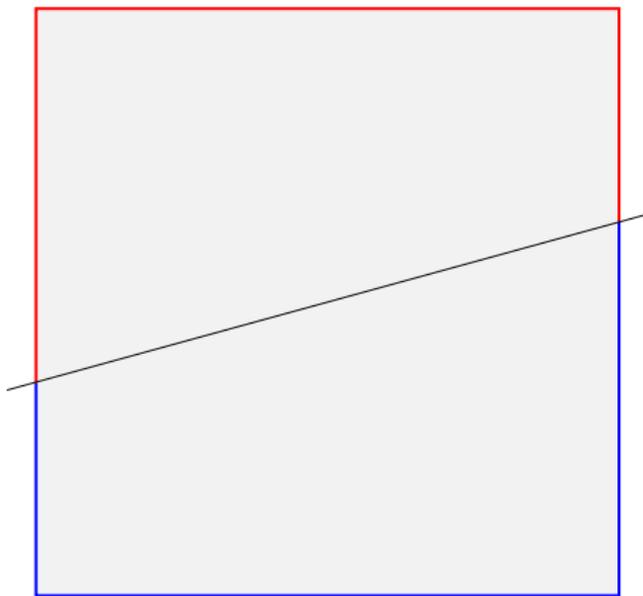
There exists $\beta_c \in (0, \infty)$ such that

- For all $\beta < \beta_c$: unique equilibrium phase
- For all $\beta > \beta_c$: q different equilibrium phases

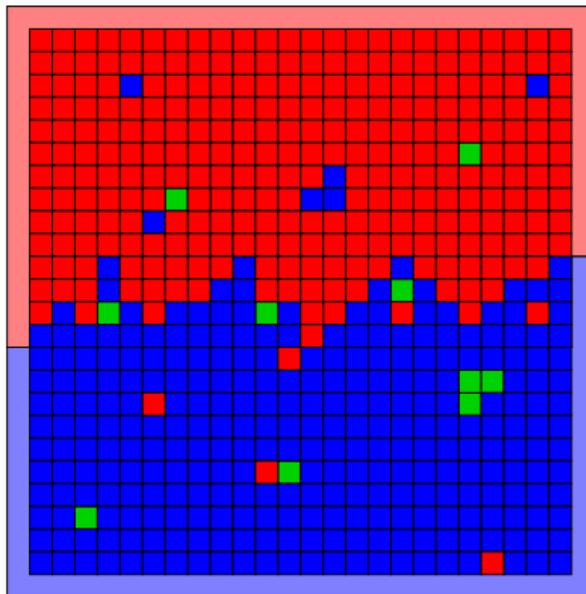
Assumption:

$$\beta > \beta_c$$

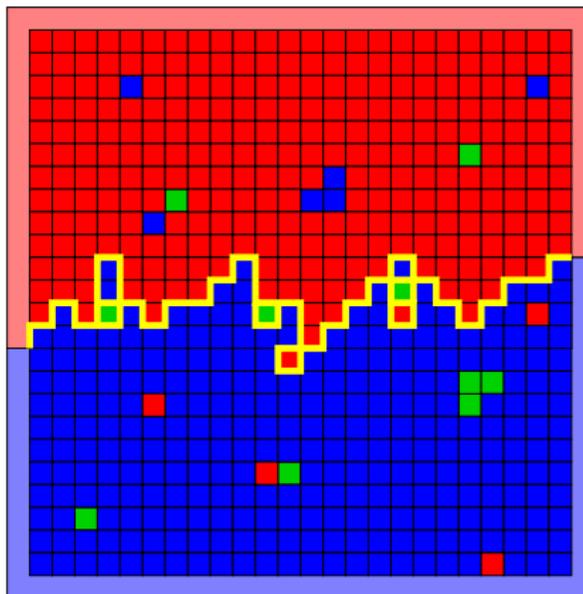
Dobrushin boundary condition



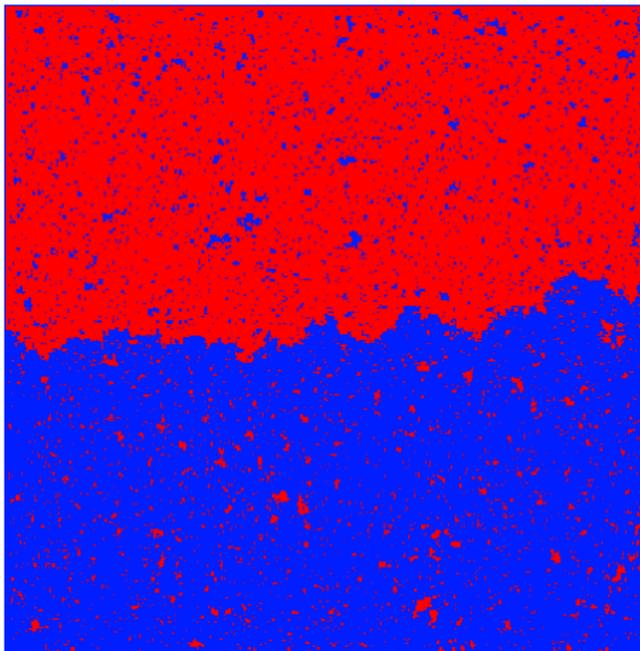
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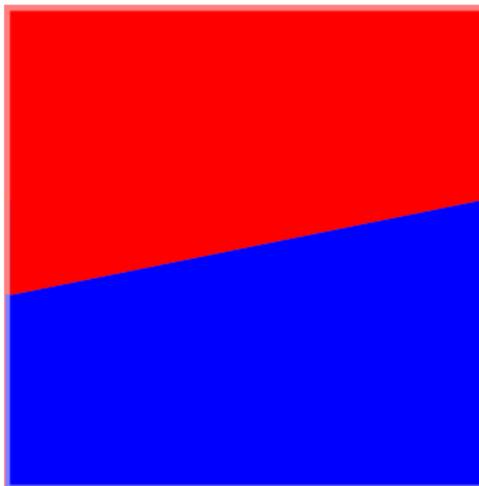


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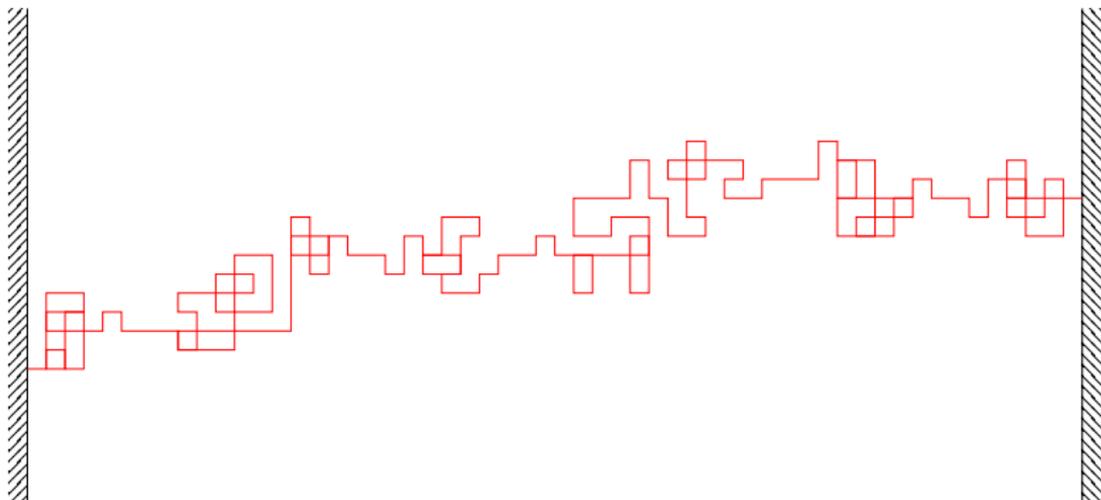
Geometry of the interface

Macro-scale: not very interesting



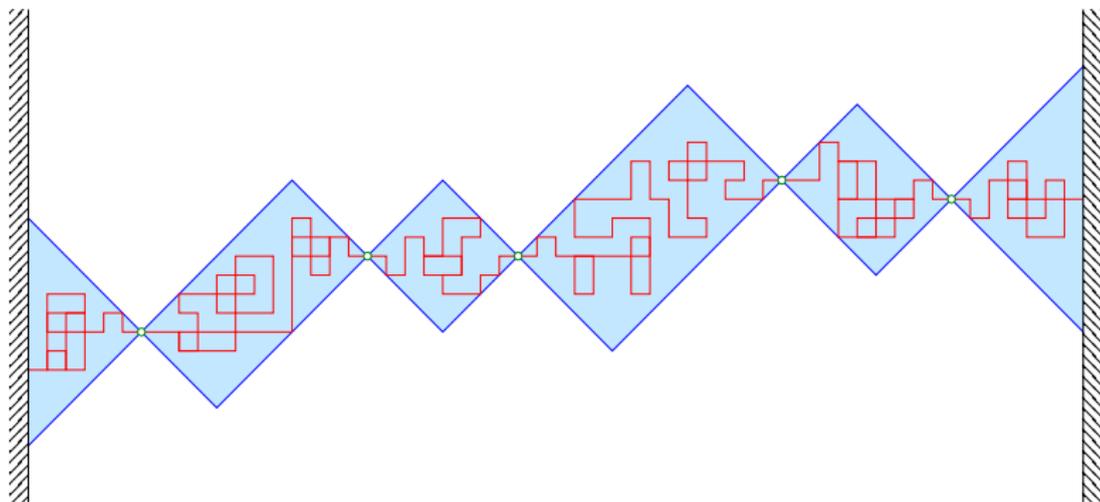
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Micro-scale: effective random walk picture



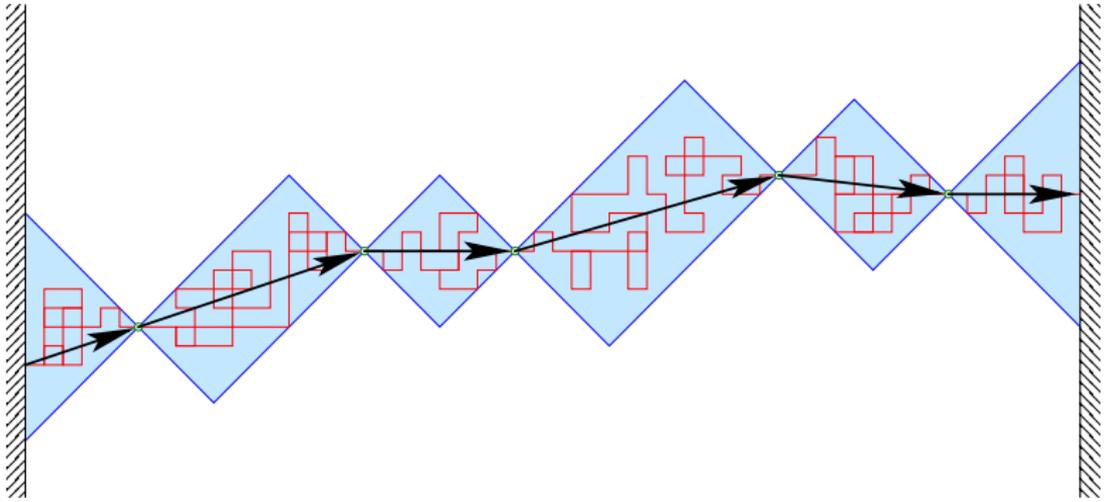
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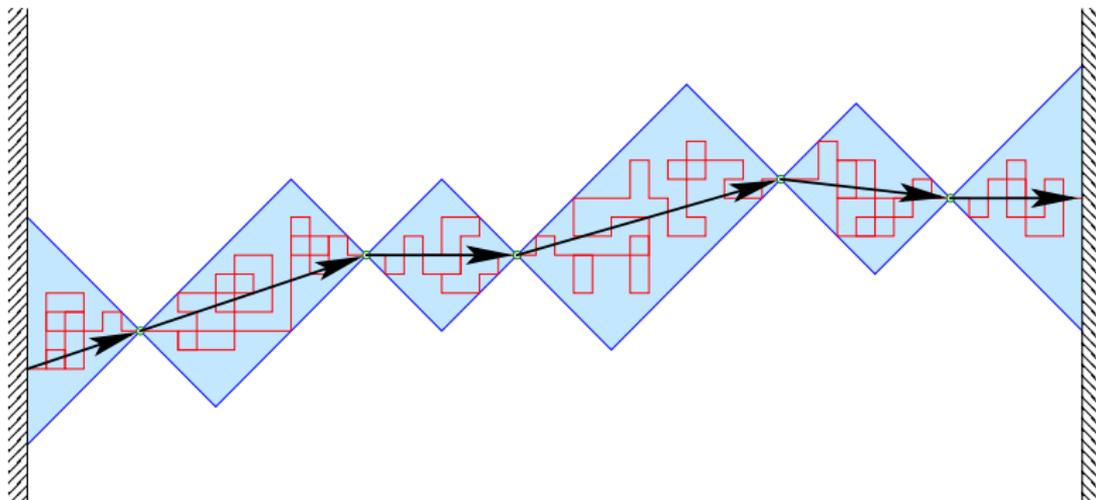
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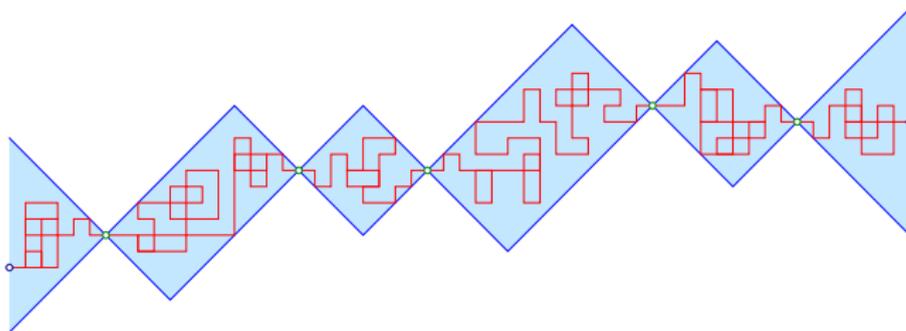
Micro-scale: effective random walk picture



(⚠ Steps are not independent!)

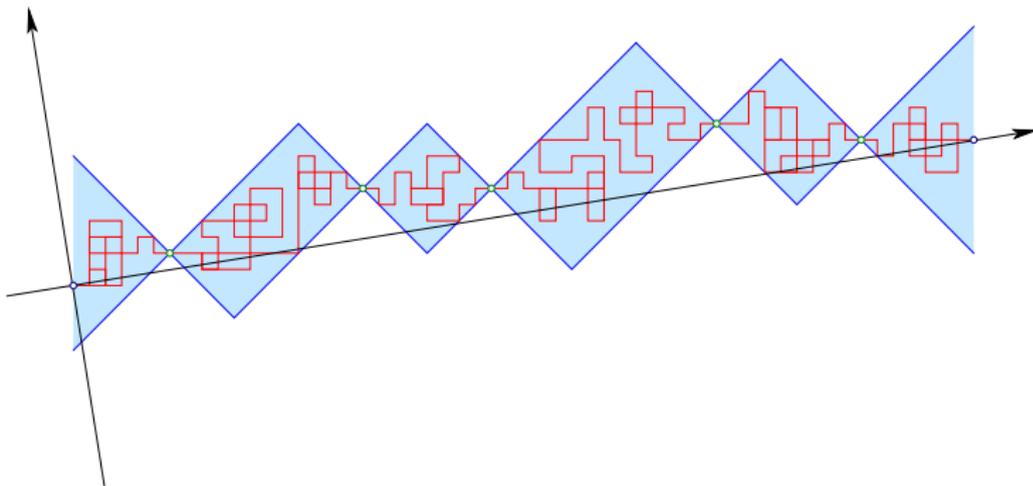
Geometry of the interface

Meso-scale: interface fluctuations



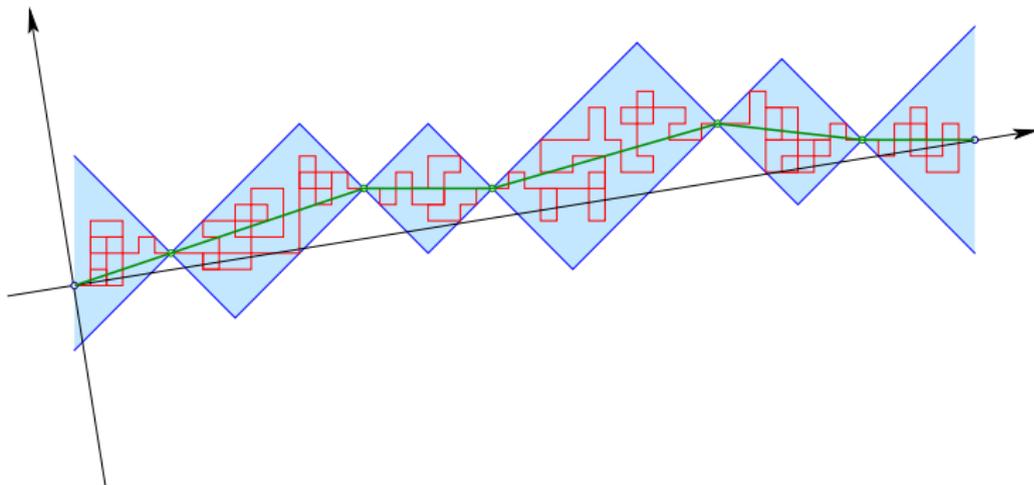
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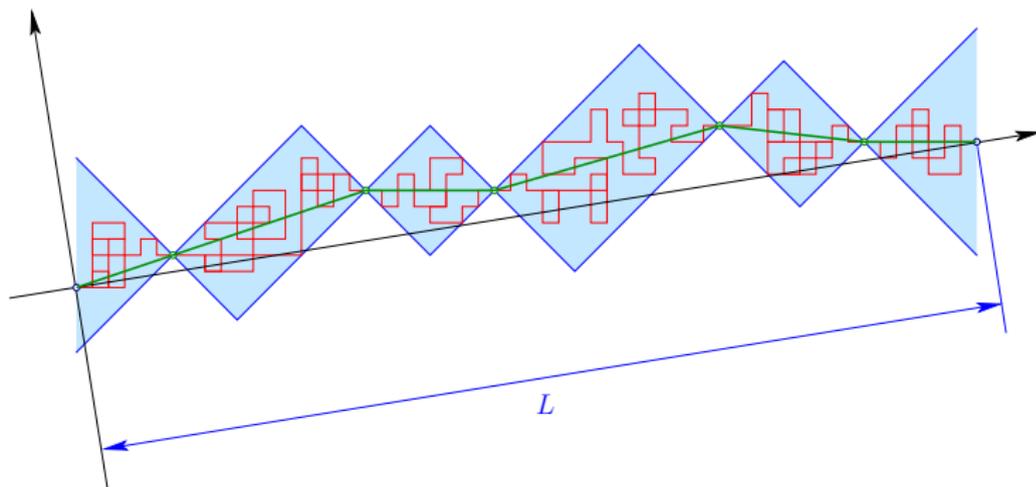
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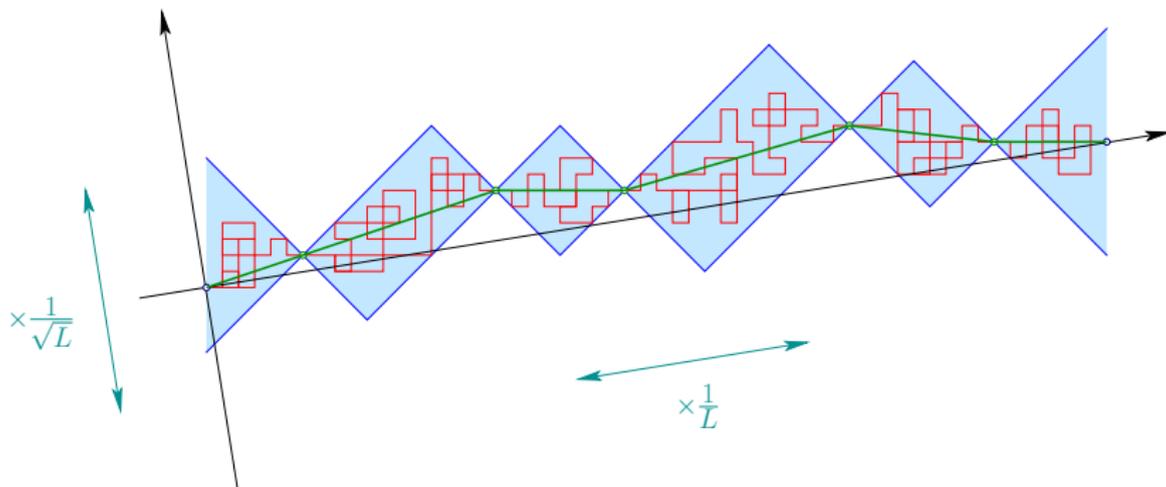
Geometry of the interface

Meso-scale: interface fluctuations



Geometry of the interface

Meso-scale: interface fluctuations



Geometry of the interface

Result

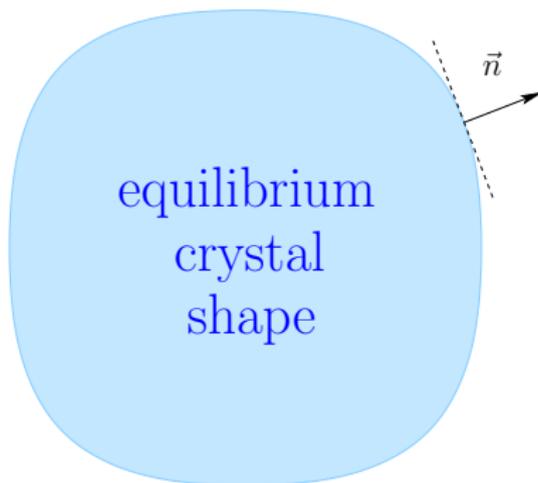
Convergence to $\sqrt{\chi_{\beta,q}(\vec{n})} B$

B : standard Brownian bridge on $[0, 1]$

$\chi_{\beta,q}(\vec{n})$: curvature of the (Wulff) equilibrium crystal shape

\vec{n} : normal to the (macroscopic) interface

Geometry of the interface



Equilibrium crystal shape: (deterministic) shape of a macroscopic droplet of one equilibrium phase immersed inside another.

Geometry of the interface

Moreover, we can deduce that the equilibrium crystal shape possesses

- an **analytic boundary**,
- a **uniformly positive curvature**.

Geometry of the interface

Moreover, we can deduce that the equilibrium crystal shape possesses

- an analytic boundary
- a uniformly positive curvature

In particular: **no roughening** in 2D Potts models.

EXAMPLE #2

Large clusters in subcritical percolative systems

FK percolation

(a.k.a. *random cluster model*)

- Introduced by Fortuin and Kasteleyn in 1972.
- 2 parameters: $p \in [0, 1]$, $q \in \mathbb{R}^+$.
- Reduces to Bernoulli percolation when $q = 1$, and to the q -states Potts model for $q = 2, 3, 4, \dots$

FK percolation

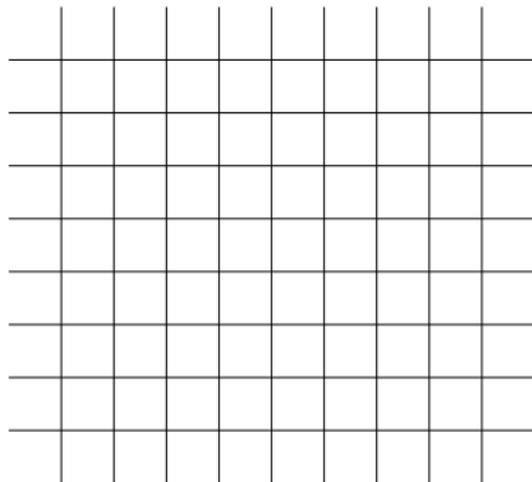
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In the sequel: $q \geq 1$.

FK percolation

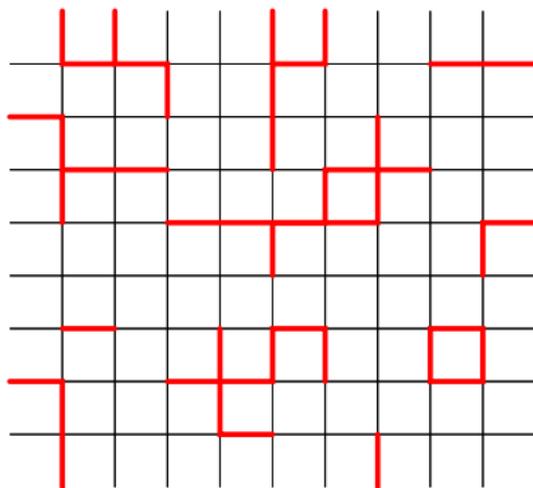
$$\Lambda \in \mathbb{E}^d$$



FK percolation

$$\Lambda \in \mathbb{E}^d$$

$$\omega \in \{0, 1\}^\Lambda$$



FK percolation

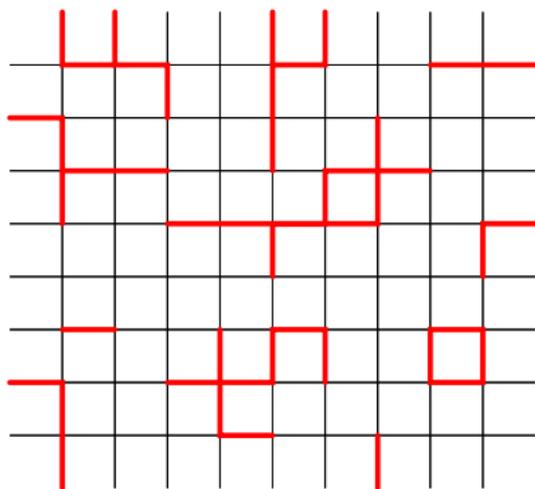
$$\Lambda \in \mathbb{E}^d$$

$$\omega \in \{0, 1\}^\Lambda$$

$$p \in [0, 1]$$

$$q \in \mathbb{R}, q \geq 1$$

$$N(\omega) = \text{number of clusters}$$



FK percolation

Infinite volume probability measure: $\mathbb{P}^{p,q}$

FK percolation

Infinite volume probability measure: $\mathbb{P}^{p,q}$

There exists $p_c(q, d) > 0$ such that

$$\mathbb{P}^{p,q}(0 \leftrightarrow \infty) = 0 \quad \forall p < p_c(q, d)$$

$$\mathbb{P}^{p,q}(0 \leftrightarrow \infty) > 0 \quad \forall p > p_c(q, d)$$

FK percolation

Infinite volume probability measure: $\mathbb{P}^{p,q}$

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$$\mathbb{P}^{p,q}(0 \leftrightarrow \infty) > 0 \quad \forall p > p_c(q, d)$$

In the sequel, we shall always assume that $p < p_c(q, d)$.
(In particular: unique infinite volume measure.)

Connectivity function

Basic quantity: connectivity function

$$\mathbb{P}^{p,q}(0 \leftrightarrow x)$$

Reduces to Potts model 2-point correlation function when
 $q = 2, 3, 4, \dots$

Connectivity function

Assumption:

$$p < p_c(q)$$

(slightly cheating here)

In particular, there exists $\xi_{p,q}(\vec{n}) > 0$ such that

$$\mathbb{P}^{p,q}(0 \leftrightarrow x) \leq e^{-\xi_{p,q}(\vec{n}_x)|x|} \quad (\vec{n}_x = x/|x|)$$

OZ behaviour

Result

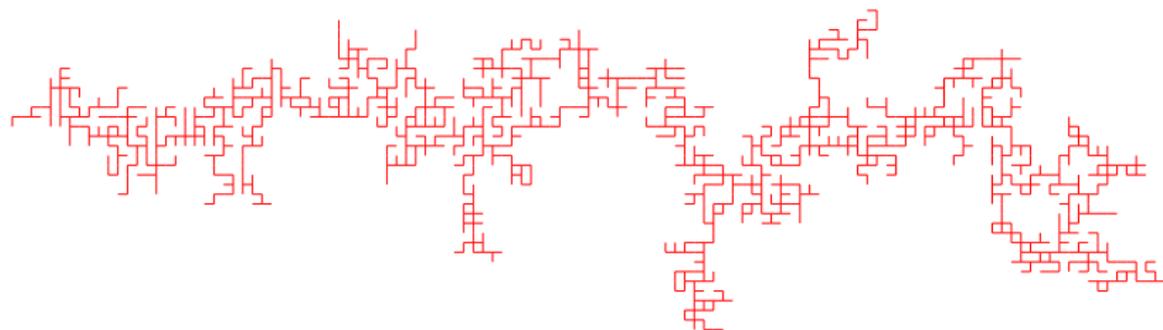
$$\mathbb{P}^{p,q}(0 \leftrightarrow x) = \frac{\Psi_{p,q}(\vec{n}_x)}{|x|^{(d-1)/2}} e^{-\xi_{p,q}(\vec{n}_x) |x|} (1 + o(1))$$

uniformly as $|x| \rightarrow \infty$. The functions $\Psi_{p,q}$ and $\xi_{p,q}$ are **positive, analytic** functions on \mathbb{S}^{d-1} .

In particular: OZ behaviour for Potts 2-point correlation functions

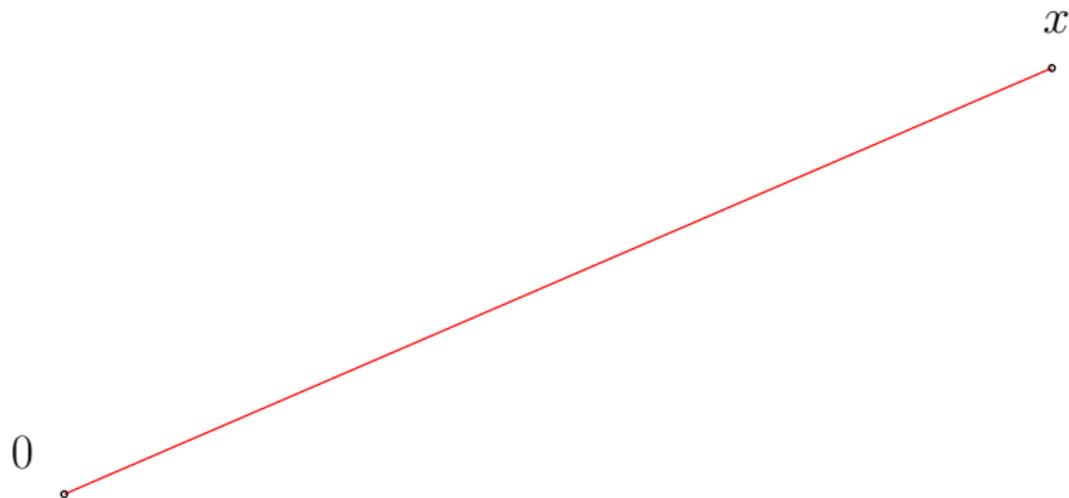
Geometry

Typical shape of large clusters under $\mathbb{P}^{p,q}(\cdot | 0 \leftrightarrow x)$ ($|x| \gg 1$)?



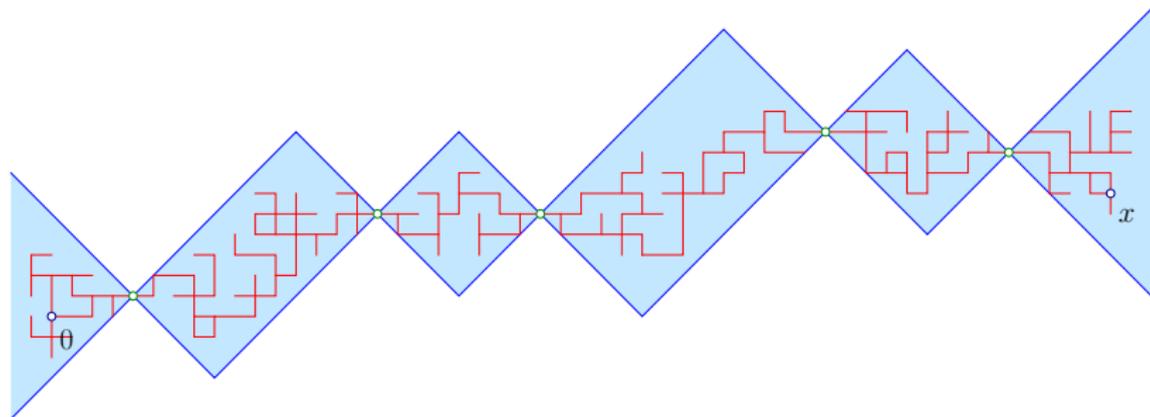
Geometry

Macro-scale: not very interesting



Geometry

Micro-scale: effective random walk picture



(⚠ Again, steps are not independent, except for $q = 1$)

Geometry

Meso-scale: fluctuations

Result

After similar scaling as before, convergence to

$$\left(\sqrt{\chi_{\beta,q}^1(x)} B^1, \dots, \sqrt{\chi_{\beta,q}^{d-1}(x)} B^{d-1} \right)$$

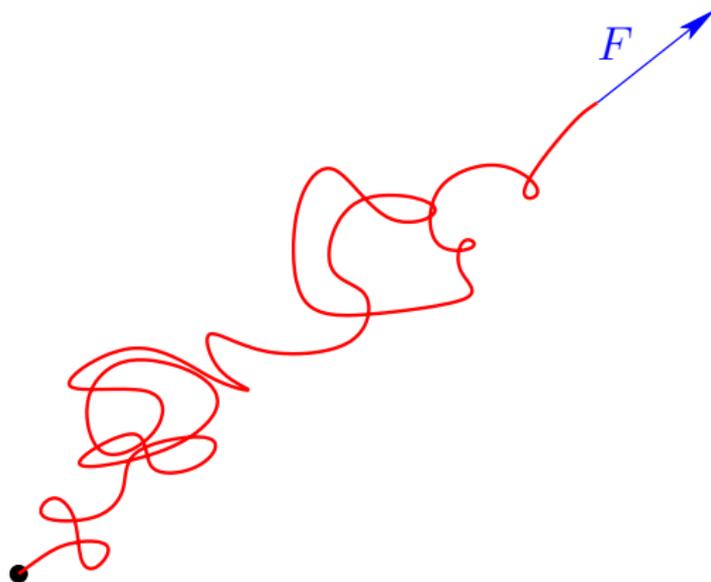
B^k : indep. standard Brownian bridges on $[0, 1]$

$\chi_{\beta,q}^k(x)$: principal curvatures of “Wulff shape” associated to $\xi_{p,q}$

EXAMPLE #3

Selfinteracting polymers

Selfinteracting polymers



Selfinteracting polymers

$$\mathbb{P}_n^F(\gamma) \propto e^{-\Phi(\gamma) + \langle F, \gamma(n) \rangle}$$

- Polymer chain: $\gamma = (\gamma(0), \dots, \gamma(n))$
(nearest-neighbour path on \mathbb{Z}^d)
- Force applied to free end: F
- Local times: $\ell_x(\gamma) = \sum_{k=0}^n \mathbf{1}_{\{\gamma(k)=x\}}$
(also possible with edges)
- Potential: $\Phi(\gamma) = \sum_{x \in \mathbb{Z}^d} \phi(\ell_x(\gamma))$
 $\phi \geq 0$, nondecreasing, $\phi(0) = 0$

Selfinteracting polymers

Possible assumptions on the interaction:

- **Repulsive:** $\phi(n + m) \geq \phi(n) + \phi(m)$

Selfinteracting polymers

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- **Attractive:** $\phi(n + m) \leq \phi(n) + \phi(m)$

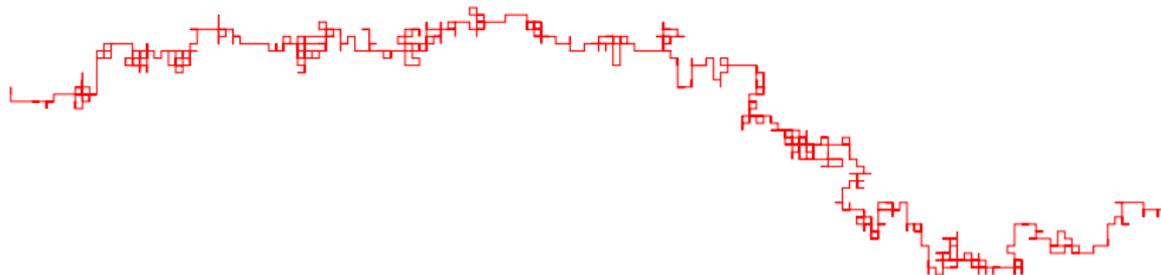
Selfinteracting polymers

Possible assumptions on the interaction:

- **Repulsive:** $\phi(n + m) \geq \phi(n) + \phi(m)$
- **Attractive:** $\phi(n + m) \leq \phi(n) + \phi(m)$
- **Small perturbations** of the pure cases, e.g.,
 - Mixed interactions (e.g., strong attr.+weak rep.)
 - Selfinteracting (e.g., reinforced) RW with drift

Geometry

Typical shape of long polymers under \mathbb{P}_n^F ?



Phase transition

Attractive case: Transition between a collapsed phase and a stretched phase.

Phase transition

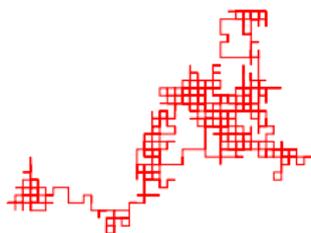
Attractive case: Transition between a collapsed phase and a stretched phase.

$\exists \mathbf{K} \subset \mathbb{R}^d$: convex set with non-empty interior s.t.

Phase transition

Attractive case: Transition between a **collapsed** phase and a stretched phase.

$\exists \mathbf{K} \subset \mathbb{R}^d$: convex set with non-empty interior s.t.

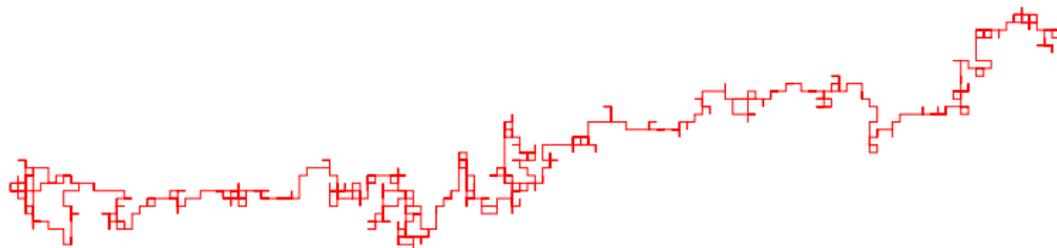


$$F \in \overset{\circ}{\mathbf{K}}$$

Phase transition

Attractive case: Transition between a collapsed phase and a **stretched** phase.

$\exists \mathbf{K} \subset \mathbb{R}^d$: convex set with non-empty interior s.t.



$$F \notin \mathbf{K}$$

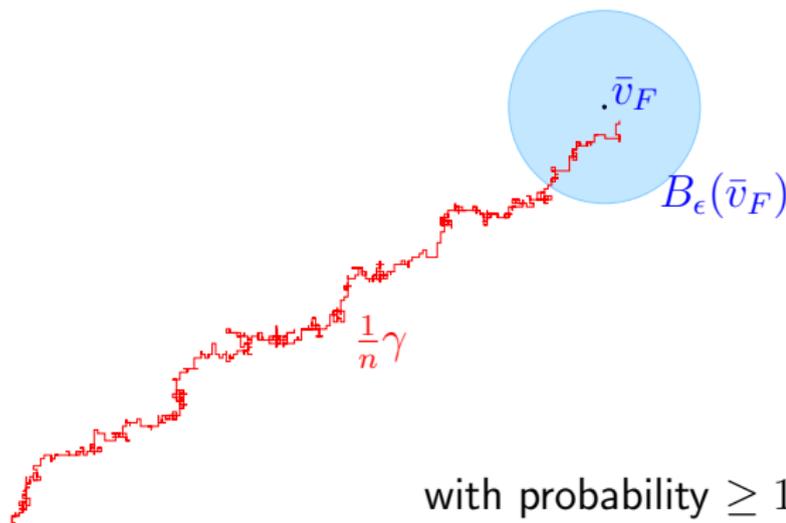
Stretched phase

The following results hold in the stretched phase, i.e. when

- $F \neq 0$ in the repulsive case
- $F \notin \mathbf{K}$ in the attractive case

Geometry of long polymers

There exists $\bar{v}_F \neq 0$ such that



Geometry of long polymers

Inside $B_\epsilon(\bar{v}_F)$:

$$\mathbb{P}_n^F \left(\frac{\gamma(n)}{n} = x \right) = \frac{G(x)}{\sqrt{n^d}} e^{-nJ_F(x)} (1 + o(1))$$

G : positive and analytic on $B_\epsilon(\bar{v}_F)$

J_F : positive, analytic on $B_\epsilon(\bar{v}_F)$, and strictly convex
with a non-degenerate quadratic minimum at \bar{v}_F

Geometry of long polymers

Moreover, the typical shape satisfies

- **macroscale**: straight line
- **microscale**: effective random walk structure
- **mesoscale**: Brownian limit

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To keep things simple, we only consider Bernoulli bond percolation on \mathbb{Z}^d at $p < p_c$.

Facts: inverse correlation length

For all $x \in \mathbb{R}^d$,

$$\xi(x) = - \lim_{k \rightarrow \infty} \frac{1}{k} \log \mathbb{P}^p(0 \leftrightarrow [kx])$$

exists and is norm on \mathbb{R}^d . Moreover,

$$\mathbb{P}^p(0 \leftrightarrow x) \leq e^{-\xi(\vec{n}_x)|x|}$$

[Menshikov '86, Aizenman, Barsky '87]

Facts: inequalities

Harris-FKG: For all A, B increasing events,

$$\mathbb{P}^p(A \cap B) \geq \mathbb{P}^p(A)\mathbb{P}^p(B)$$

BK: For all A, B increasing events,

$$\mathbb{P}^p(A \circ B) \leq \mathbb{P}^p(A)\mathbb{P}^p(B)$$

where \circ denotes disjoint occurrence.

Equidecay set and Wulff shape

Two convex bodies are naturally associated to ξ :

Equidecay set and Wulff shape

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Equidecay set

$$U_\xi = \{x \in \mathbb{R}^d : \xi(x) \leq 1\}$$

Equidecay set and Wulff shape

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Wulff shape

$$K_\xi = \{t \in \mathbb{R}^d : (t, \vec{n})_d \leq \xi(\vec{n}), \forall \vec{n} \in \mathbb{S}^{d-1}\}$$

Equidecay set and Wulff shape

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Equidecay set

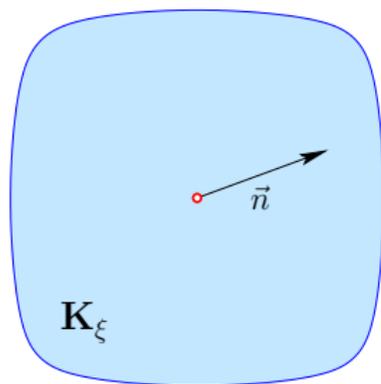
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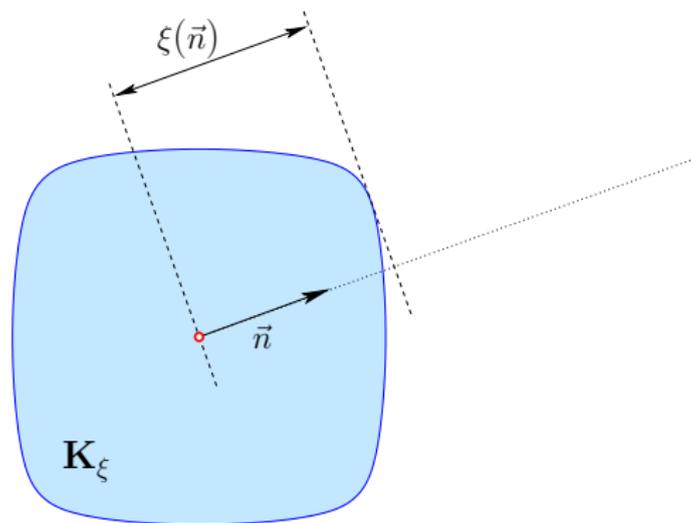
Each set encodes all the information about ξ .

Equidecay set and Wulff shape



$$K_\xi = \{t \in \mathbb{R}^d : (t, \vec{n})_d \leq \xi(\vec{n}), \forall \vec{n} \in \mathbb{S}^{d-1}\}$$

Equidecay set and Wulff shape



$$\mathbf{K}_\xi = \{t \in \mathbb{R}^d : (t, \vec{n})_d \leq \xi(\vec{n}), \forall \vec{n} \in \mathbb{S}^{d-1}\}$$

Equidecay set and Wulff shape

\mathbf{U}_ξ and \mathbf{K}_ξ are **polar**:

$$\mathbf{U}_\xi = \left\{ x \in \mathbb{R}^d : \max_{t \in \mathbf{K}_\xi} (t, x)_d \leq 1 \right\}$$

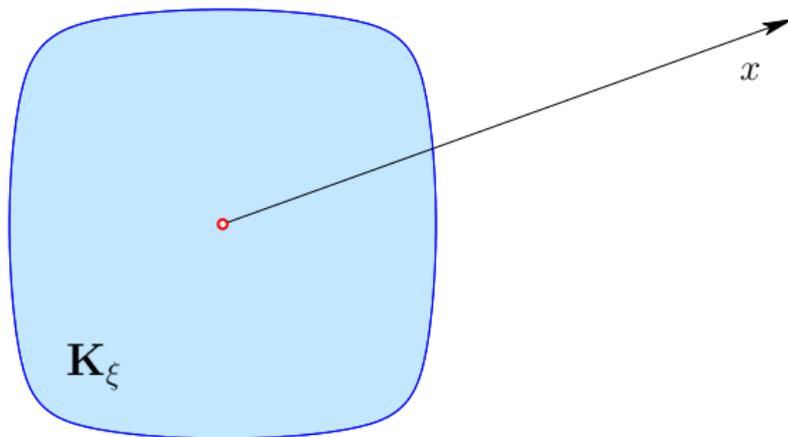
$$\mathbf{K}_\xi = \left\{ t \in \mathbb{R}^d : \max_{x \in \mathbf{U}_\xi} (t, x)_d \leq 1 \right\}$$

Equidecay set and Wulff shape

$x \in \mathbb{R}^d$ and $t \in \partial\mathbf{K}_\xi$ are **dual** if $(t, x)_d = \xi(x)$.

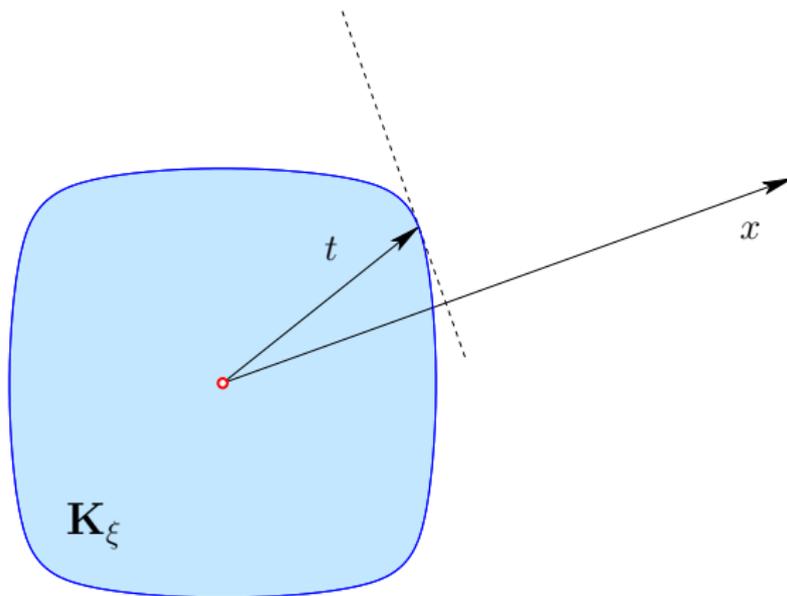
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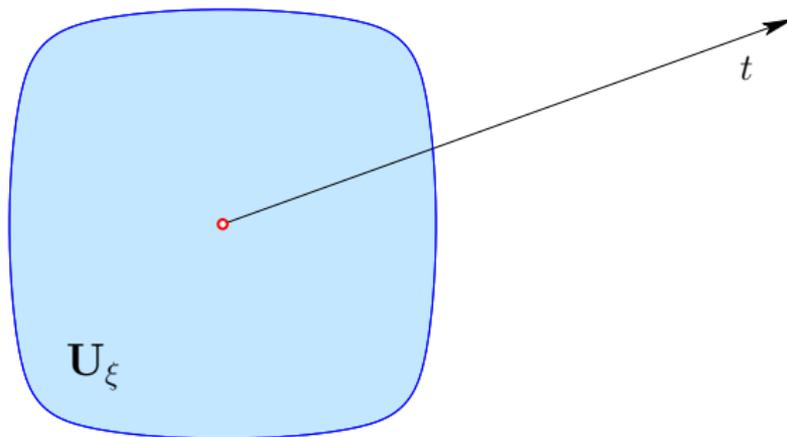
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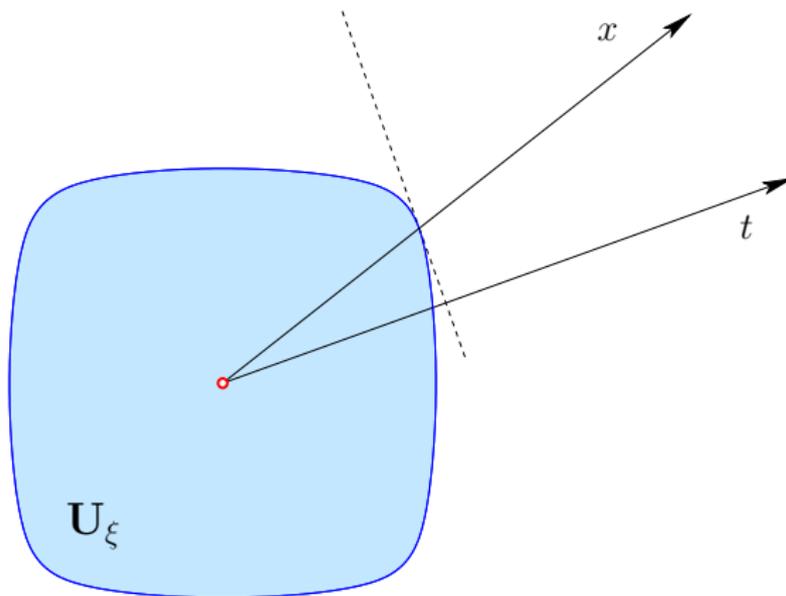
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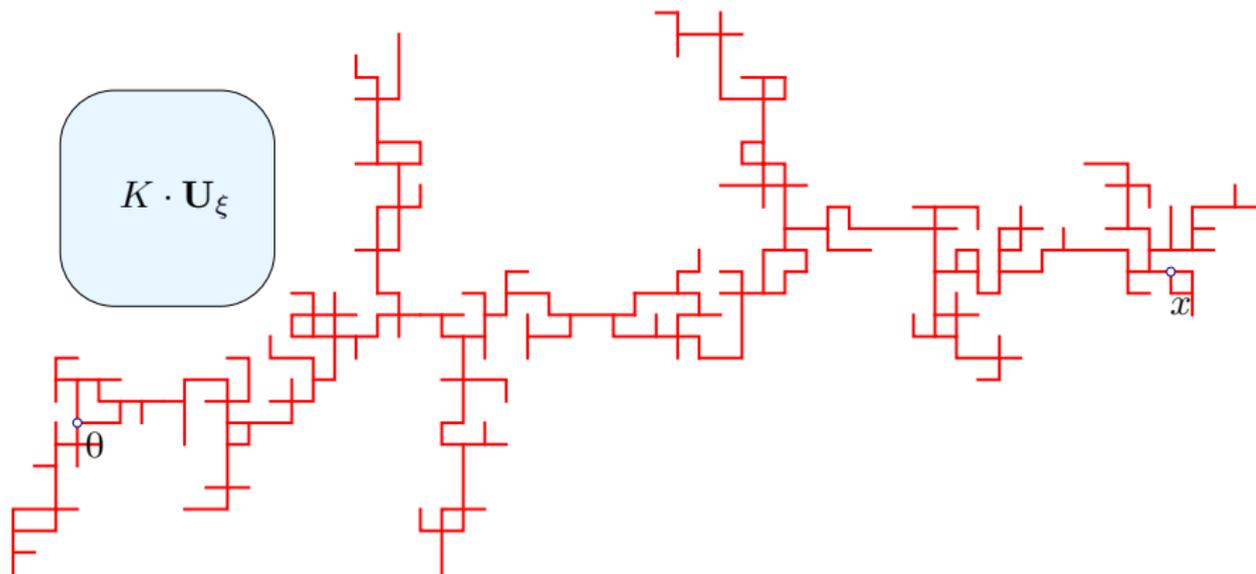


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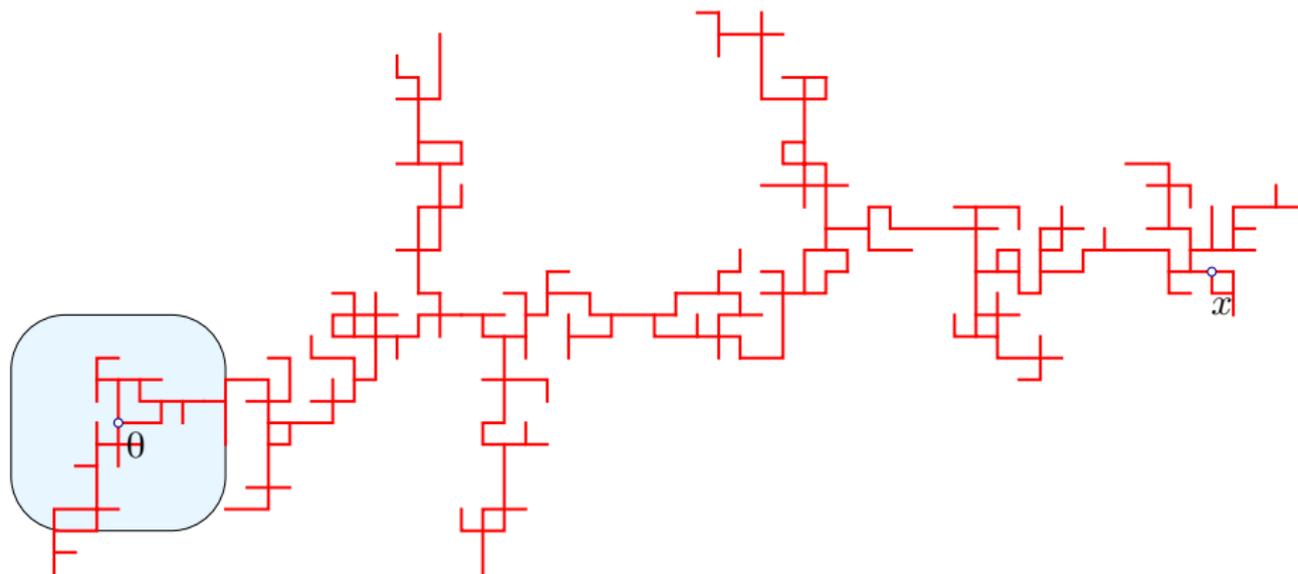
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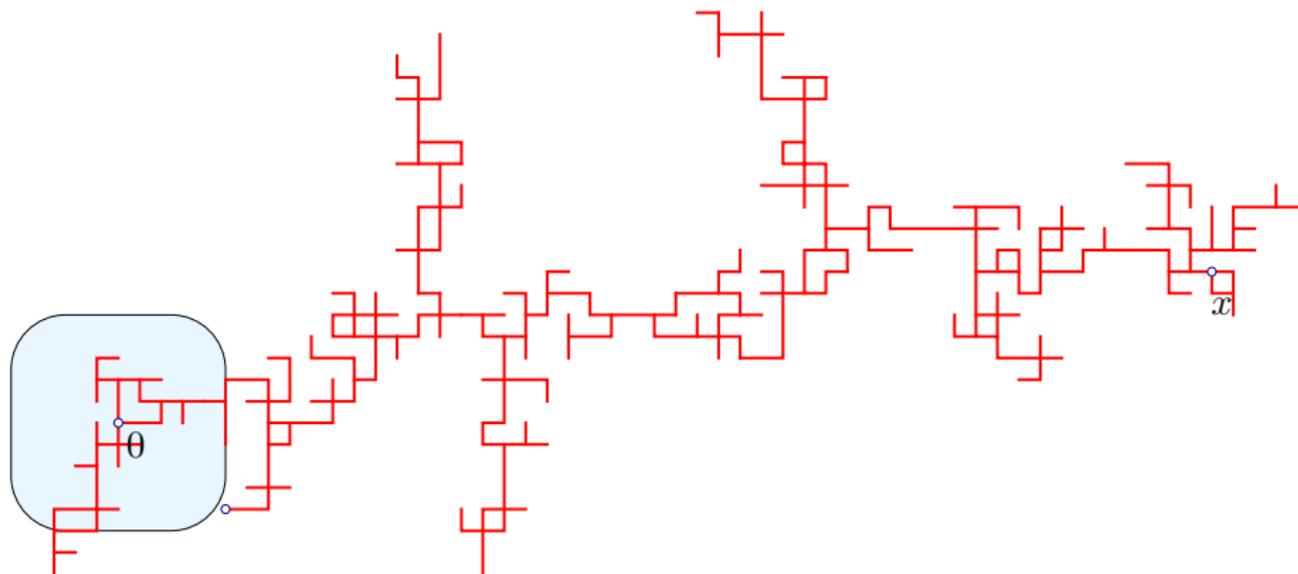
Skeleton



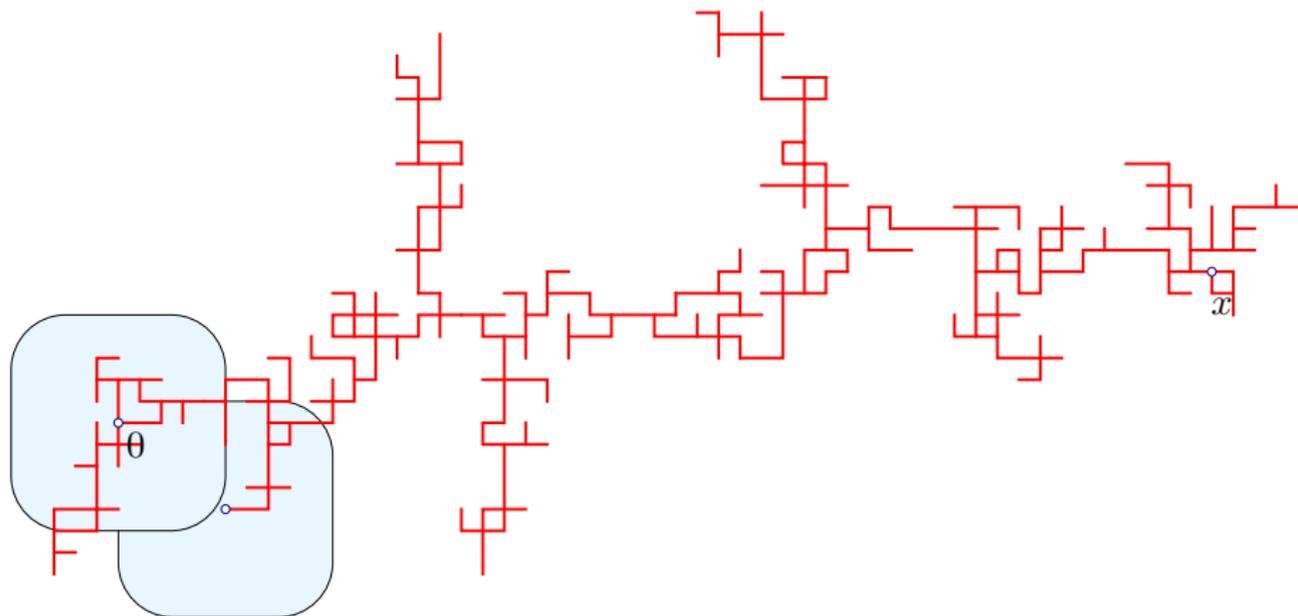
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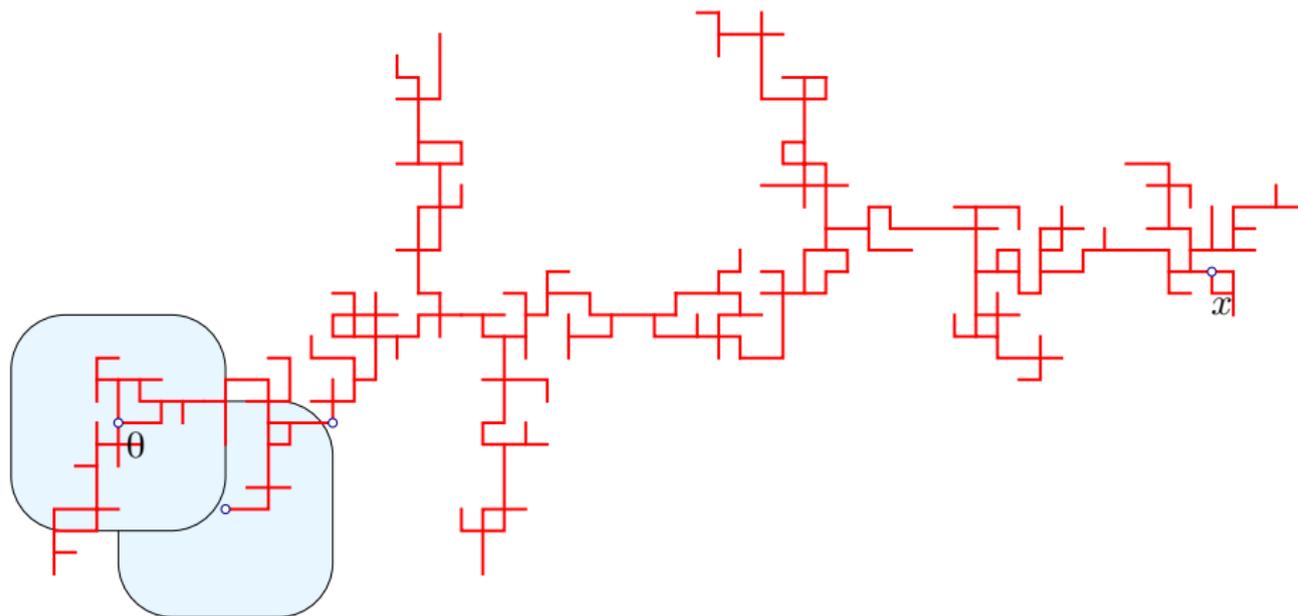
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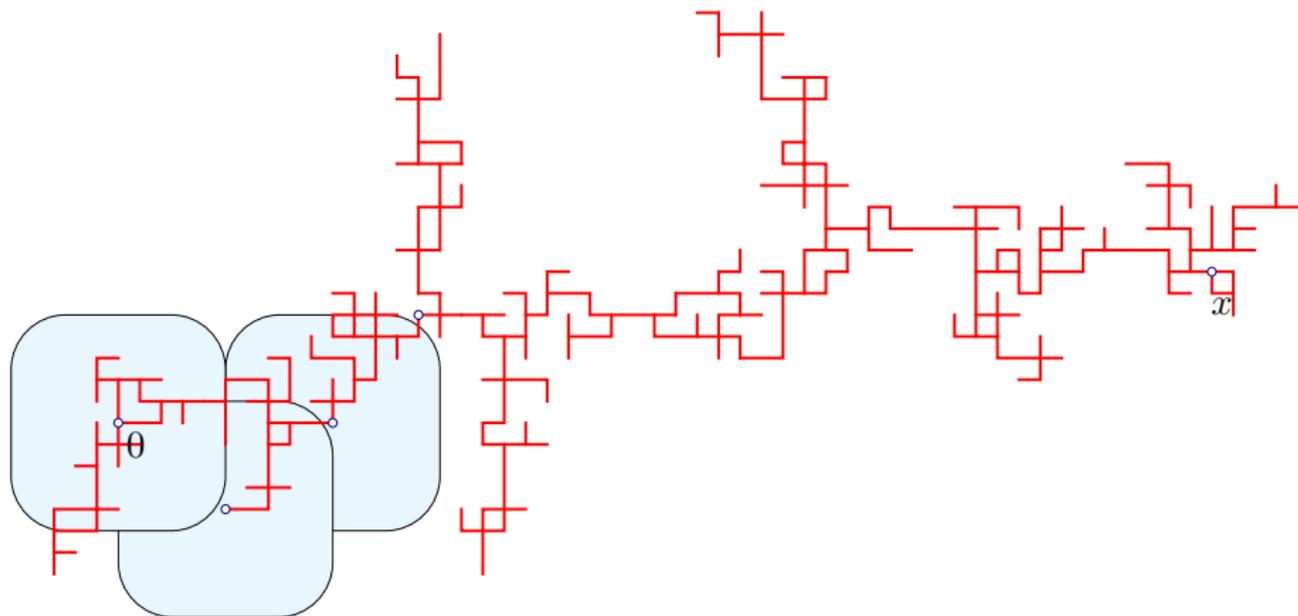
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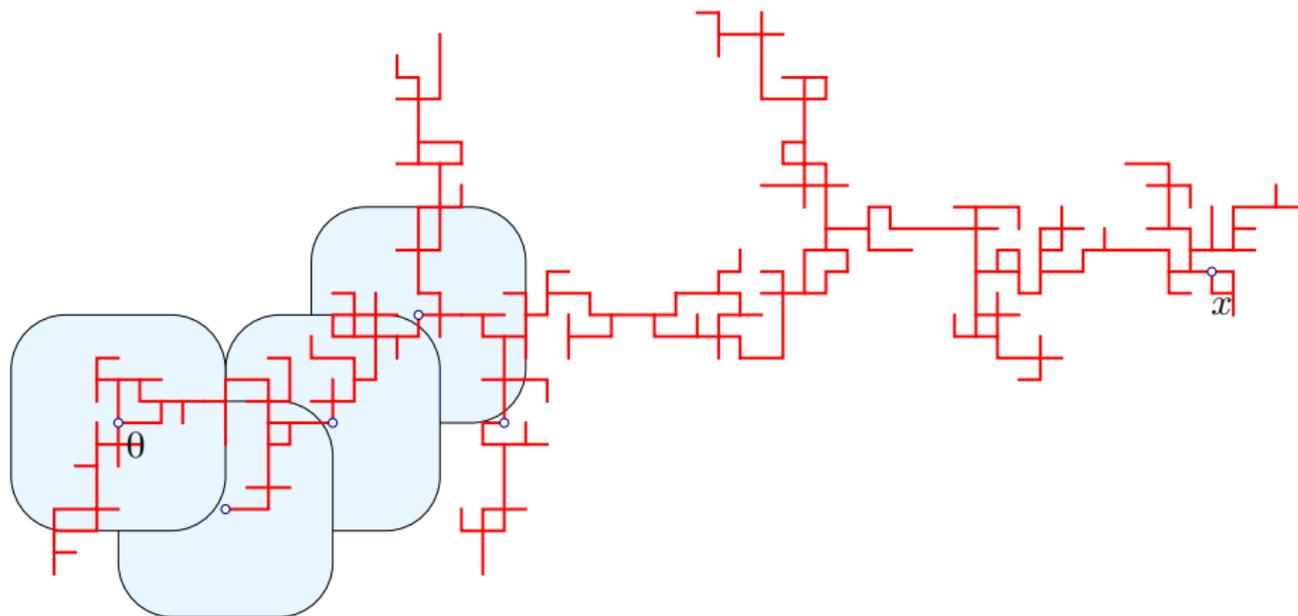
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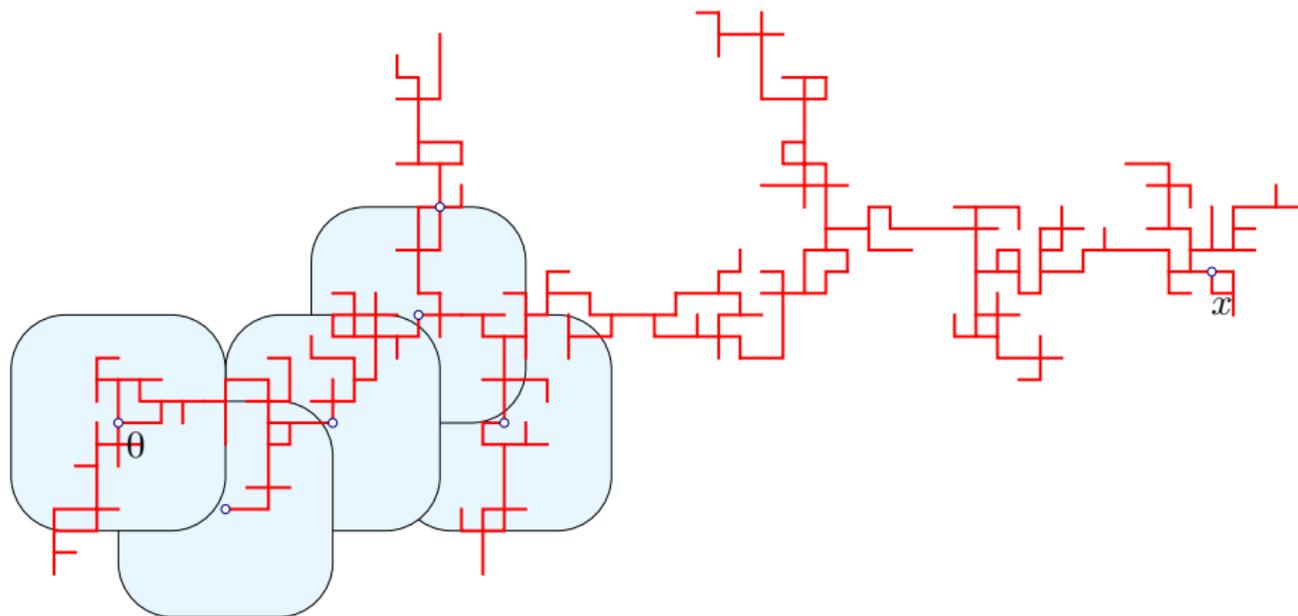
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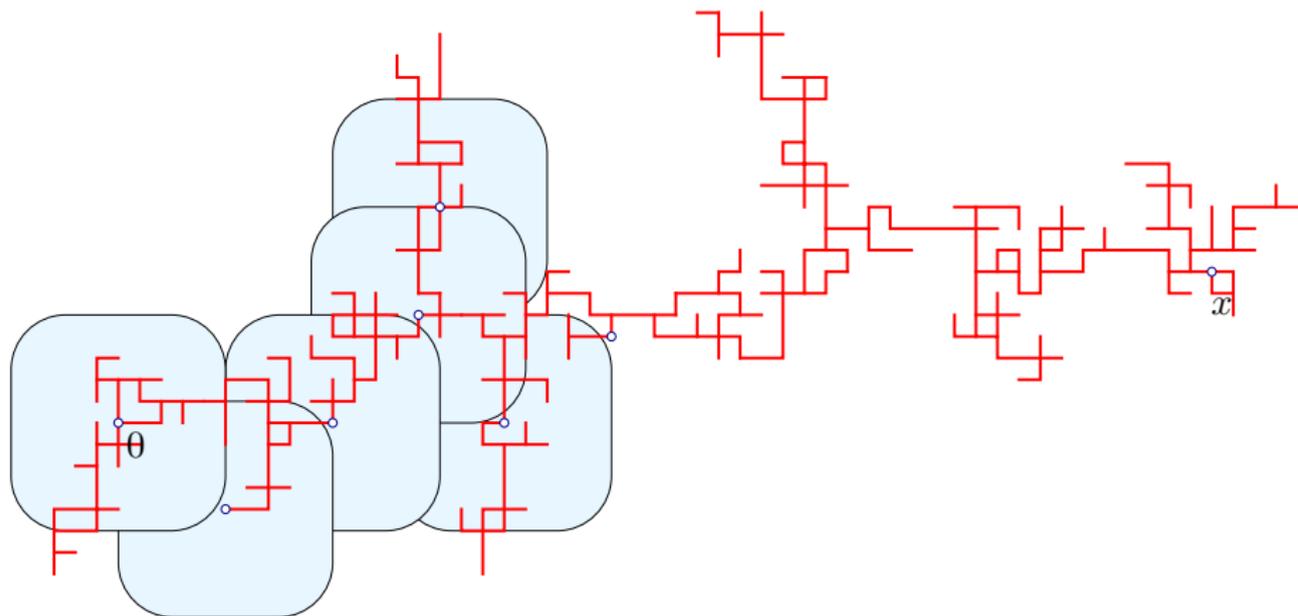
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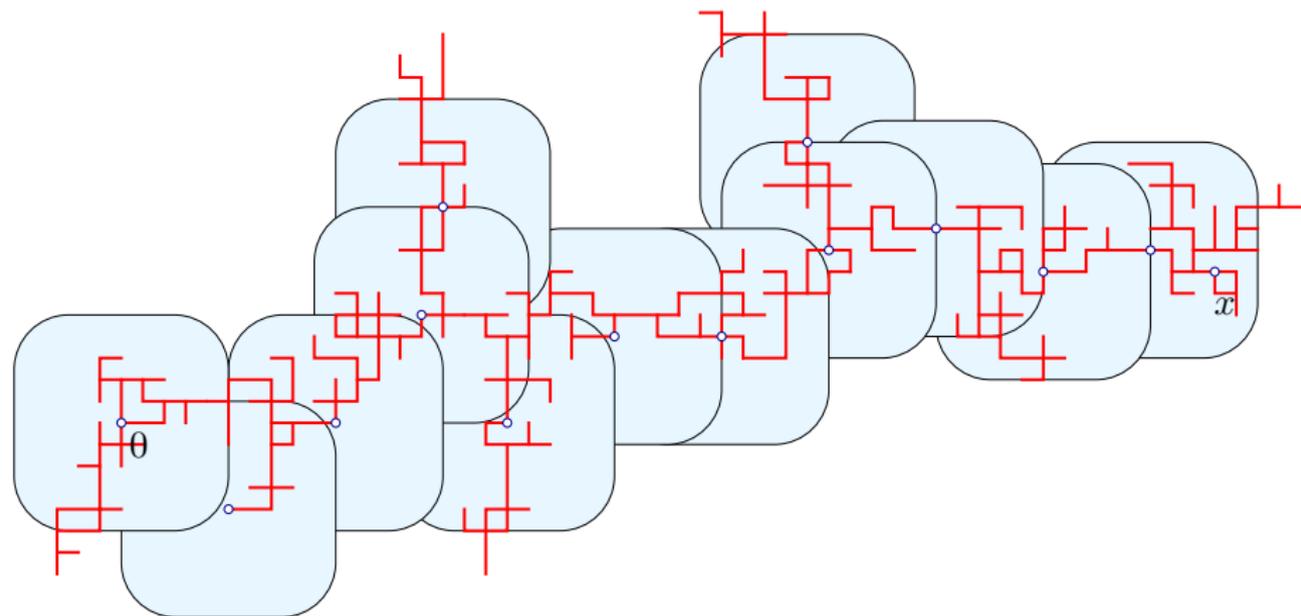
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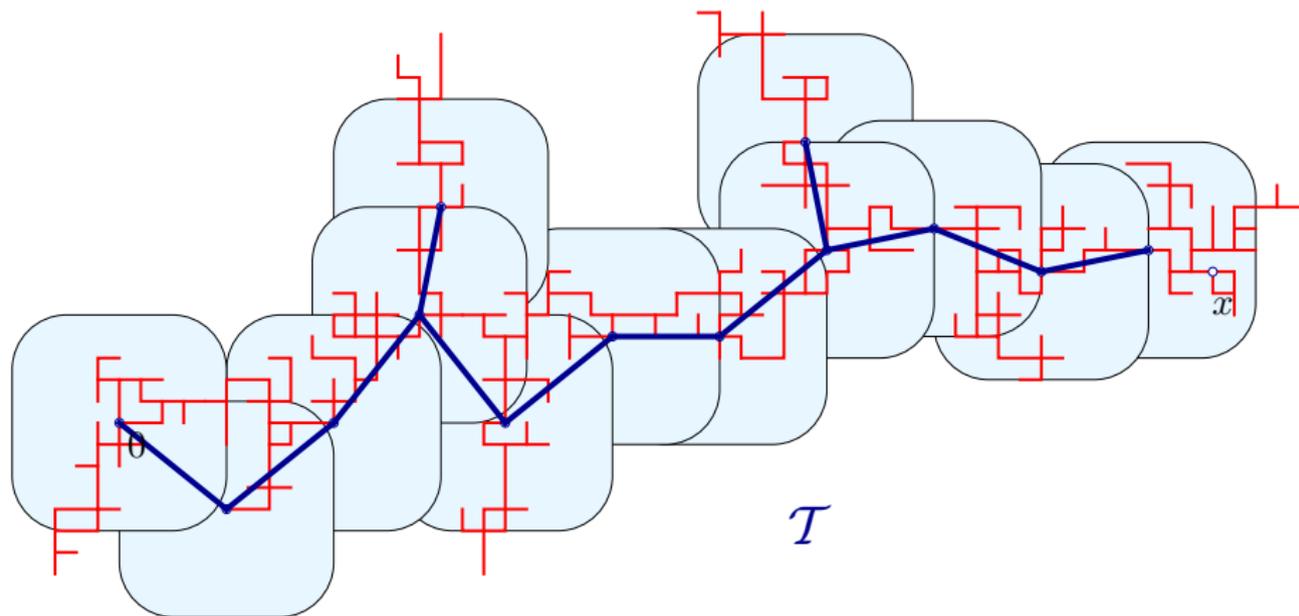
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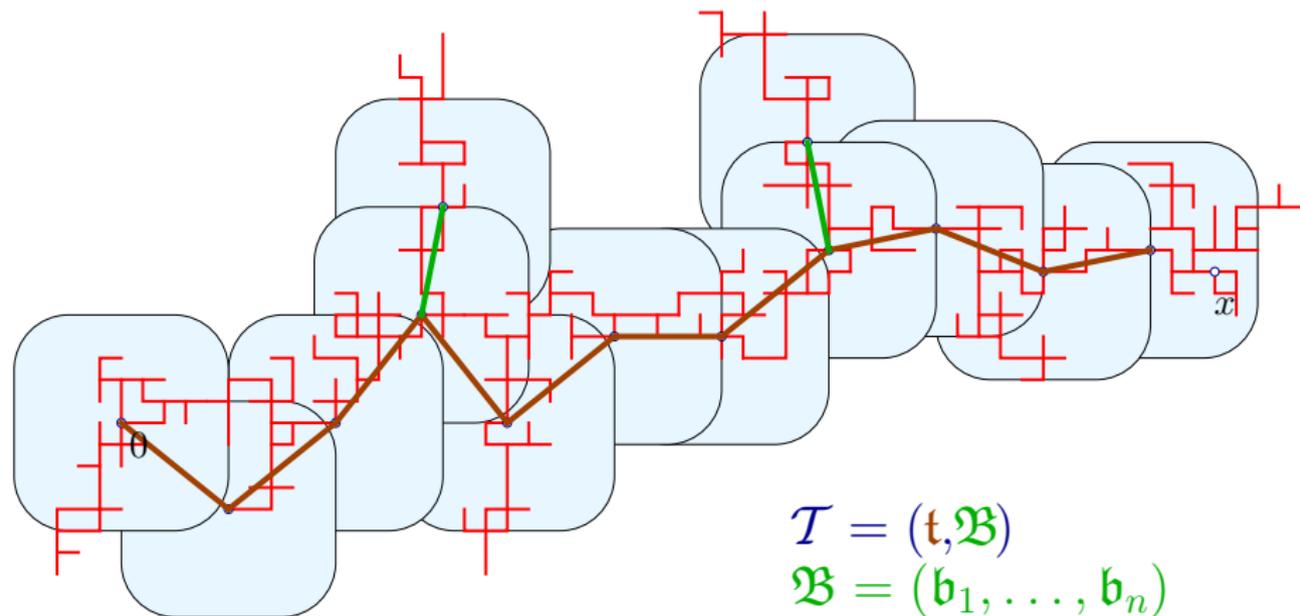
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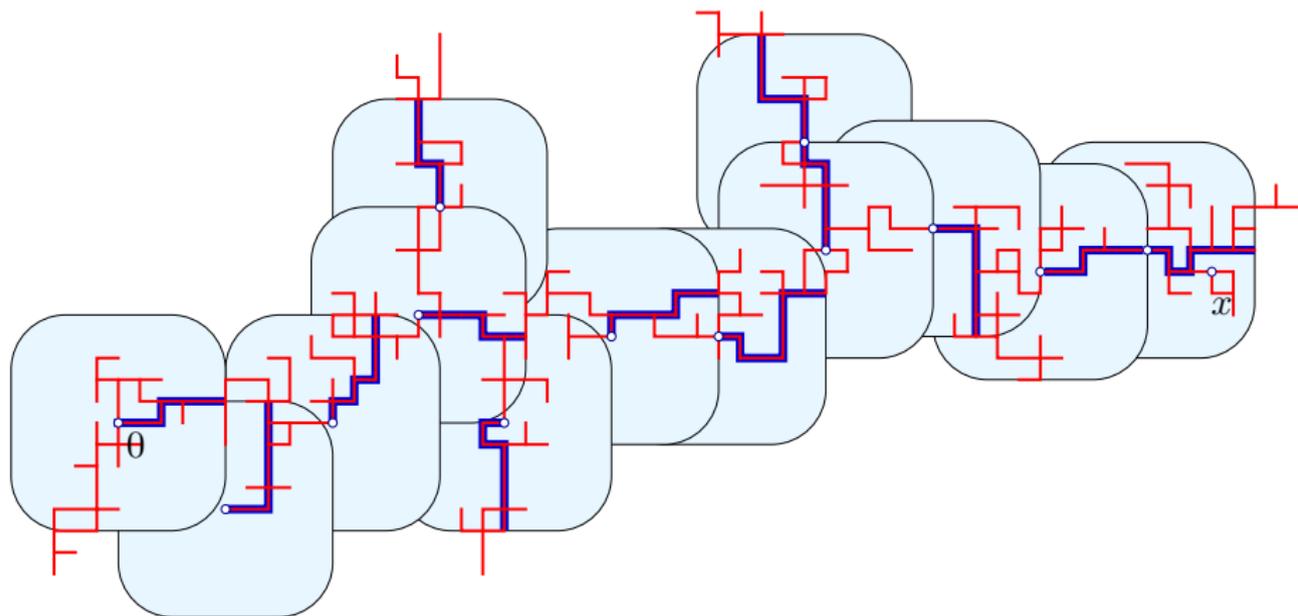
Skeleton



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Rough bounds

Let $\mathcal{T} = (0 = x_1, x_2, \dots, x_{N_{\mathcal{T}}})$, $\mathbf{U}_{\xi}^K(x_i) = x_i + K\mathbf{U}_{\xi}$ and

$$A_i = \{x_i \leftrightarrow \partial\mathbf{U}_{\xi}^K(x_i)\}$$

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BK implies that

$$\begin{aligned}\mathbb{P}^{\mathbb{P}}(\mathcal{T}) &\leq \mathbb{P}^{\mathbb{P}}(A_1 \circ A_2 \circ \dots \circ A_{N_{\mathcal{T}}}) \\ &\leq \prod_{i=1}^{N_{\mathcal{T}}} \mathbb{P}^{\mathbb{P}}(A_i) = \mathbb{P}^{\mathbb{P}}(0 \leftrightarrow \partial\mathbf{U}_{\xi}^K(0))^{N_{\mathcal{T}}} \\ &\leq (cK^{d-1}e^{-K})^{N_{\mathcal{T}}} = e^{-KN_{\mathcal{T}}(1+o_K(1))}\end{aligned}$$

Rough bounds

$N_{\mathfrak{t}} = \#$ of vertices in \mathfrak{t} , $N_{\mathfrak{B}} = \#$ of vertices in \mathfrak{B} .

- Typical trees have a small trunk:

$$\exists c_1, c_2 : \mathbb{P}^p(N_{\mathfrak{t}} > c_1 \frac{|x|}{K} \mid 0 \leftrightarrow x) \leq e^{-c_2|x|}$$

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- Total size of branches of typical trees is small:

$$\forall c_3 > 0 : \mathbb{P}^p(N_{\mathfrak{B}} > c_3 \frac{|x|}{K} \mid 0 \leftrightarrow x) \leq e^{-\frac{1}{2}c_3|x|}$$

Rough bounds

of trunks of size $N \lesssim (K^{d-1})^N = e^{N(d-1) \log K}$.

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Thus (with K large enough)

$$\mathbb{P}^p(N_t \geq c_1 \frac{|x|}{K}) \leq \sum_{N \geq c_1 \frac{|x|}{K}} e^{-KN(1-o_K(1))} \leq e^{-\frac{1}{2}c_1|x|}.$$

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Conclusion follows (taking c_1 large enough) since

$$\mathbb{P}^p(0 \leftrightarrow x) \geq e^{-\xi(\vec{n}_x)|x|(1+o(1))}.$$

Surcharge function

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measures the typicality of an increment.

For a trunk $\mathbf{t} = (t_0, \dots, t_{N_t})$, we set

$$\mathfrak{s}_t(\mathbf{t}) = \sum_{l=1}^{N_t} \mathfrak{s}_t(t_l - t_{l-1})$$

Surcharge function

Surcharge inequality

Let $\epsilon > 0$. There exists $K_0(\epsilon)$ such that, for all $K > K_0$,

$$\mathbb{P}(\mathfrak{s}_t(\mathbf{t}) > 2\epsilon|x| \mid 0 \leftrightarrow x) \leq e^{-\epsilon|x|}$$

uniformly in $x \in \mathbb{Z}^d$, $t \in \partial\mathbf{K}_\xi$ dual to x .

Surcharge function

$$\mathbb{P}^p(\mathbf{t}) \leq e^{-N_t K(1-o_K(1))} \leq e^{-\sum_{i=1}^{N_t} \xi(\mathbf{t}_i - \mathbf{t}_{i-1}) + o_K(1)|x|}.$$

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Now,

$$\begin{aligned} \sum_{i=1}^{N_t} \xi(\mathbf{t}_i - \mathbf{t}_{i-1}) &= \sum_{i=1}^{N_t} (\mathfrak{s}_t(\mathbf{t}_i - \mathbf{t}_{i-1}) + (t, \mathbf{t}_i - \mathbf{t}_{i-1})_d) \\ &= \mathfrak{s}_t(\mathbf{t}) + (t, \mathbf{t}_{N_t})_d \\ &= \mathfrak{s}_t(\mathbf{t}) + (t, x)_d - (t, x - \mathbf{t}_{N_t})_d \\ &= \mathfrak{s}_t(\mathbf{t}) + \xi(x) - (t, x - \mathbf{t}_{N_t})_d. \end{aligned}$$

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$$\implies \mathbb{P}^p(\mathbf{t}) \leq e^{-\mathfrak{s}_t(\mathbf{t}) - \xi(x) + o_K(1)|x|}.$$

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We can assume that $N_t \leq c_1|x|/K$.

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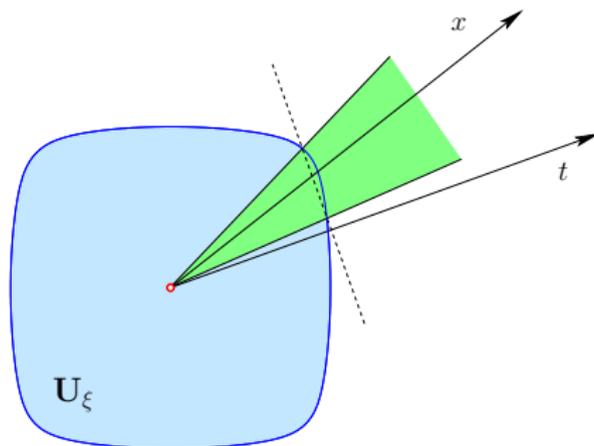
$$e^{c(\log K/K)|x|} = e^{o_K(1)|x|}.$$

Since, $\mathbb{P}^p(\mathbf{t}) \leq e^{-\mathfrak{s}_t(\mathbf{t}) - \xi(x) + o_K(1)|x|}$,

$$\mathbb{P}^p(\mathfrak{s}_t(\mathbf{t}) \geq 2\epsilon|x|) \leq e^{-(2\epsilon - o_K(1))|x|} e^{-\xi(x)},$$

and the conclusion follows as before... □

Forward cone

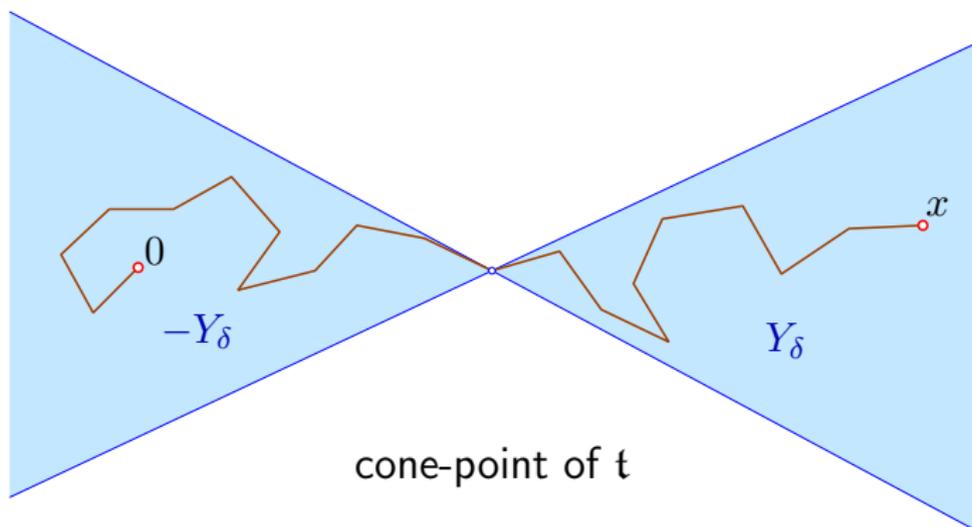


$$\begin{aligned} Y_\delta(t) &= \{y \in \mathbb{R}^d : (y, t)_d > (1 - \delta)\xi(y)\} \\ &= \{y \in \mathbb{R}^d : \mathfrak{s}_t(y) < \delta\xi(y)\} \end{aligned}$$

Cone points of trunks



Cone points of trunks



Cone points of trunks

$$\#^{\text{n.c.p.}}(\mathbf{t}) = \#\{\text{non-cone-points of } \mathbf{t}\}$$

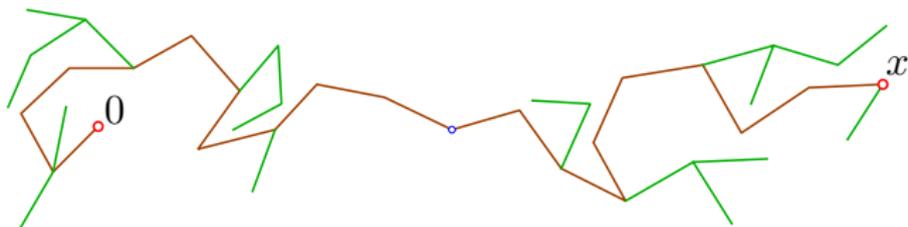
Lemma

$$\mathfrak{s}_t(\mathbf{t}) \geq c_4 \delta K \#^{\text{n.c.p.}}(\mathbf{t})$$

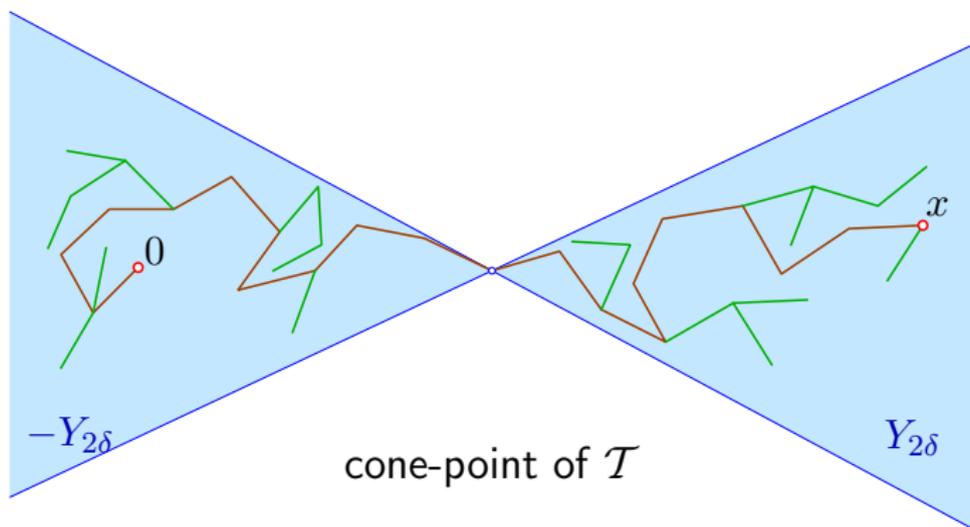
Consequently,

$$\mathbb{P}(\#^{\text{n.c.p.}}(\mathbf{t}) \geq \epsilon N(\mathbf{t}) \mid 0 \leftrightarrow x) \leq e^{-c_5 \epsilon |x|}$$

Cone points of trees



Cone points of trees

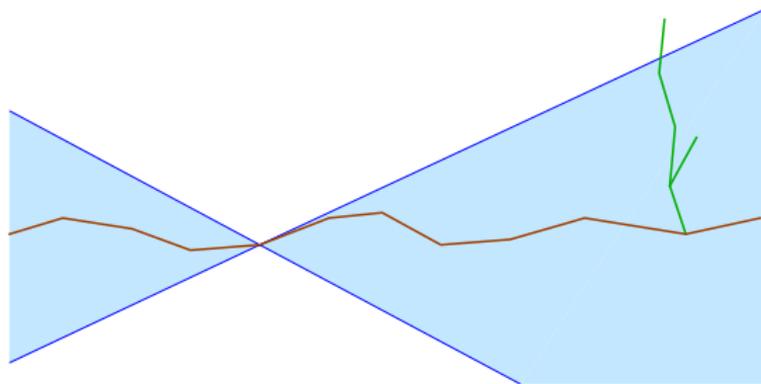


Cone points of trees

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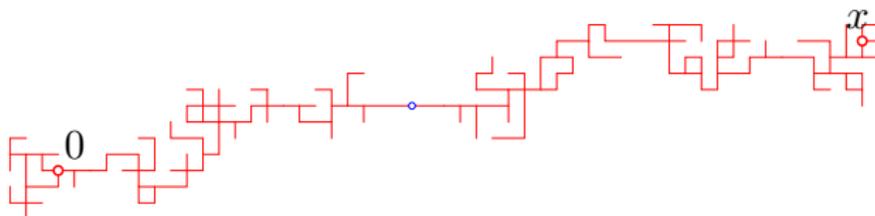
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Lemma

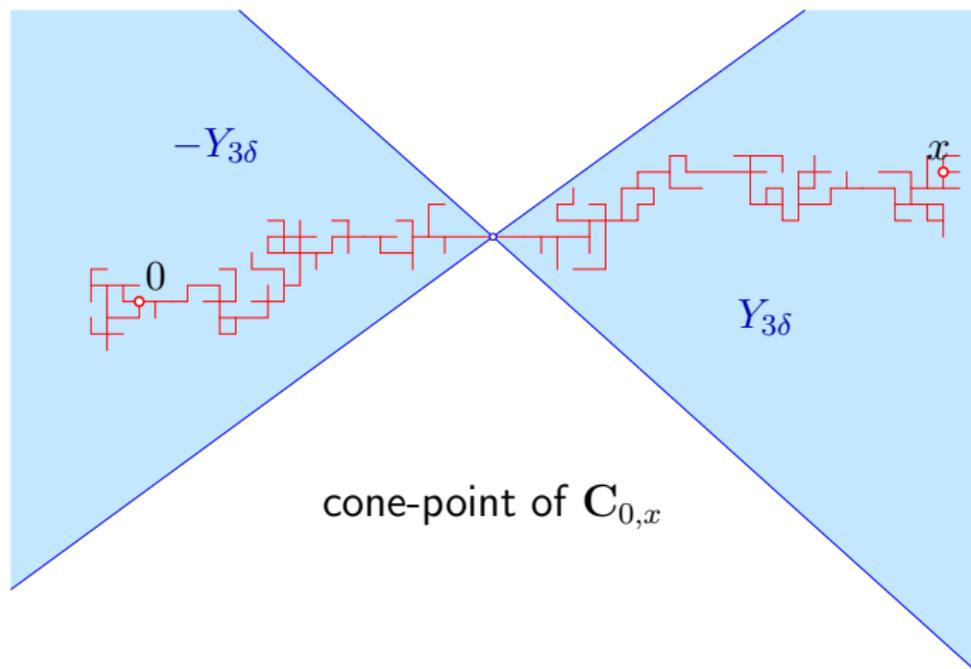
There exist $\nu > 0$ and c such that

$$\mathbb{P}(\#\{\text{cone-points of } \mathcal{T}\} < \nu \frac{|x|}{K} \mid 0 \leftrightarrow x) \leq e^{-c|x|}$$

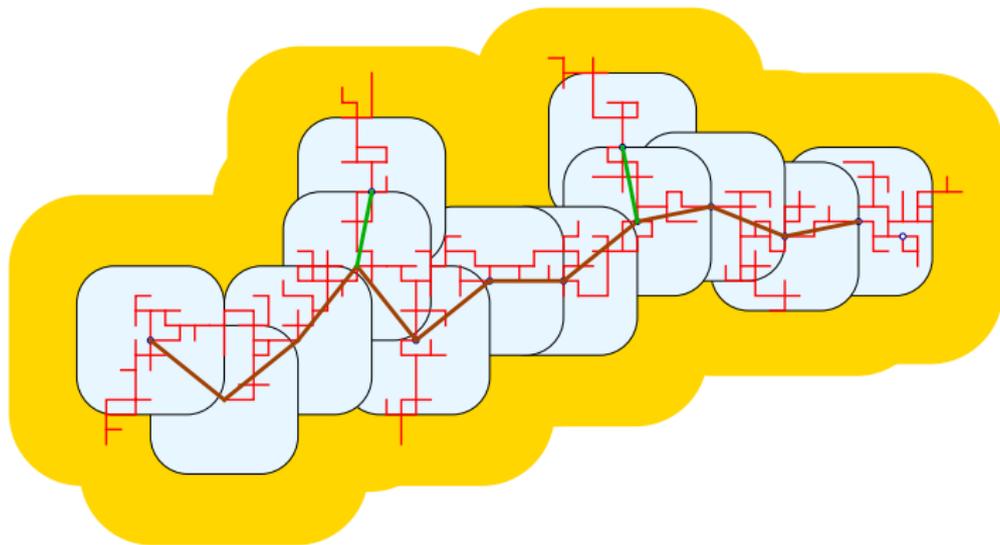
Cone points of clusters



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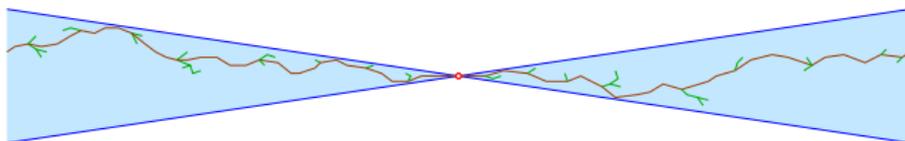


Clusters remain close to their approximating tree

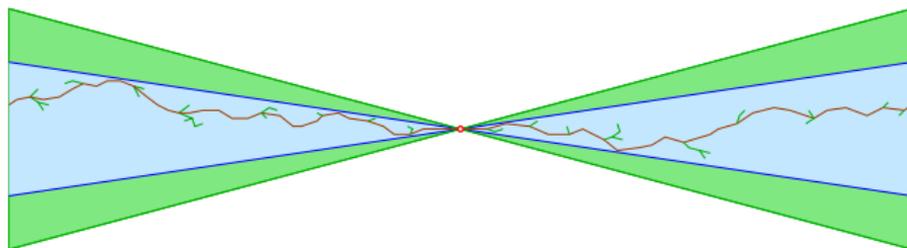
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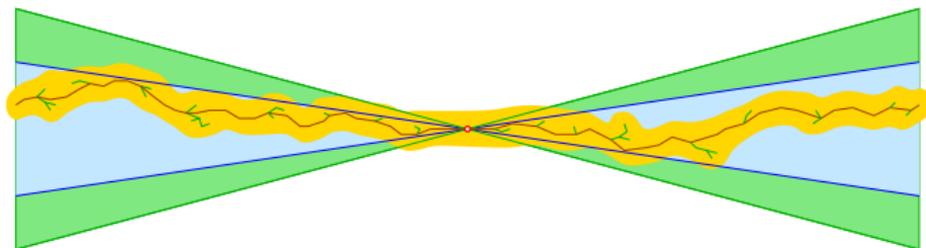
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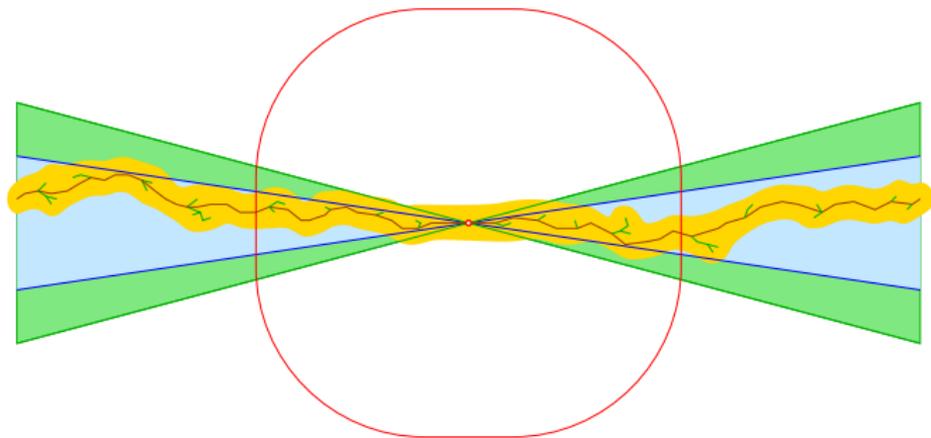
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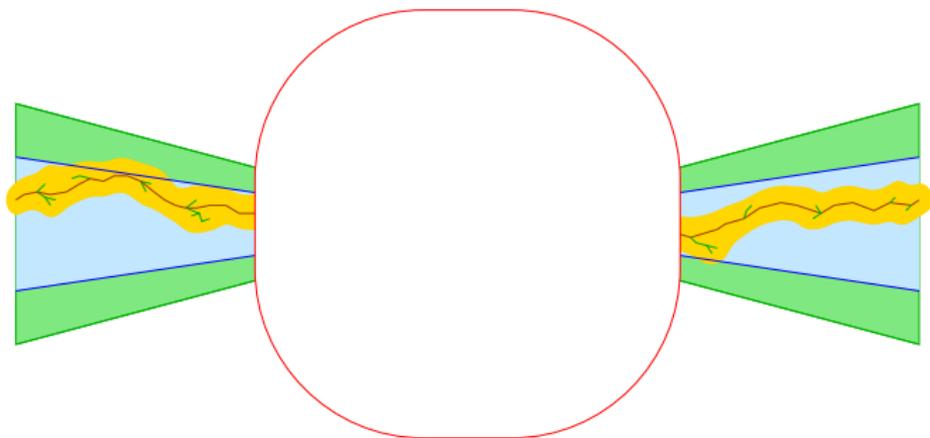
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- Up to exponentially small error, a positive density of the cone-points of \mathcal{T} are also cone-points of $\mathbf{C}_{0,x}$

Cone points of clusters

$\#_{t,\delta}^{\text{cone}}(\mathbf{C}_{0,x})$: number of cone-points of $\mathbf{C}_{0,x}$

Theorem

There exist $\delta \in (0, \frac{1}{2})$, ν and c such that

$$\mathbb{P}(\#_{t,\delta}^{\text{cone}}(\mathbf{C}_{0,x}) \leq \nu|x| \mid 0 \leftrightarrow x) \leq e^{-c|x|}$$

uniformly in x and dual t .

Decomposition into irreducible pieces

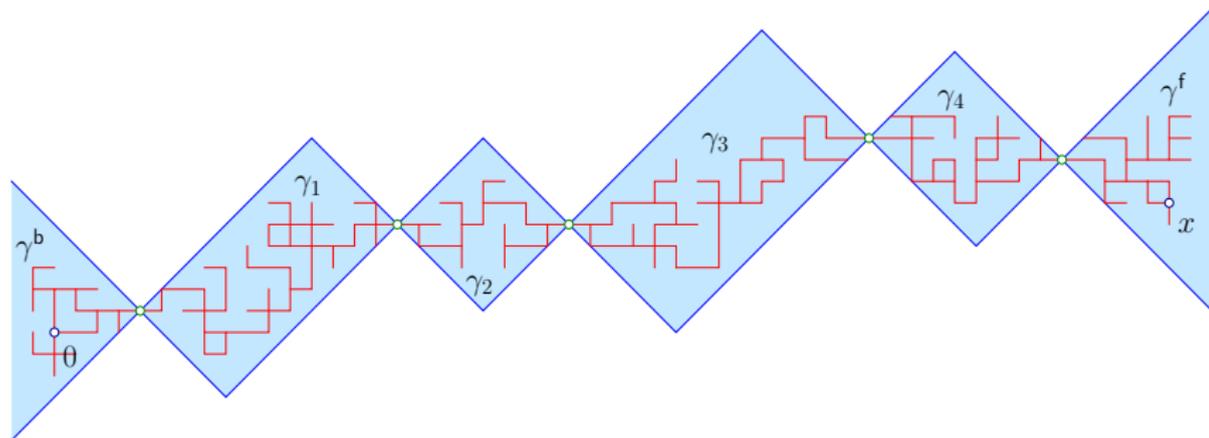
We can thus decompose the cluster $\mathbf{C}_{0,x}$ into irreducible pieces:

$$\mathbf{C}_{0,x} = \gamma^b \amalg \gamma_1 \amalg \dots \amalg \gamma_n \amalg \gamma^f$$

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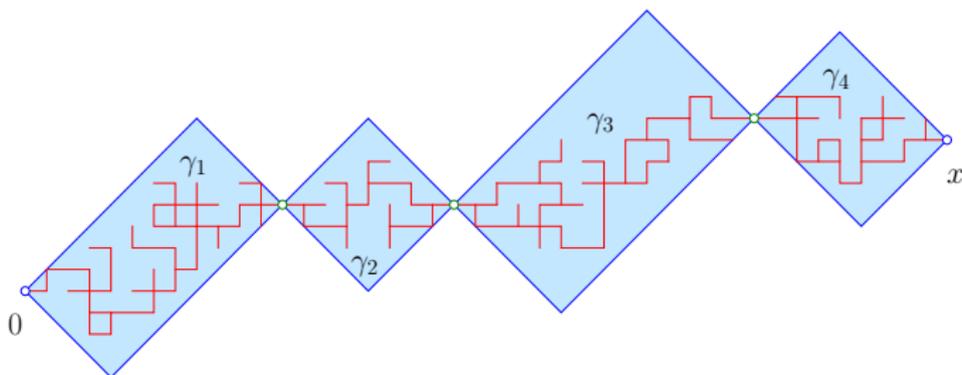
$$C_{0,x} = \gamma^b \amalg \gamma_1 \amalg \dots \amalg \gamma_n \amalg \gamma^f$$



Decomposition into irreducible pieces

To simplify, I shall assume in the sequel that:

$$C_{0,x} = \gamma_1 \amalg \dots \amalg \gamma_n$$



Decomposition into irreducible pieces

We can thus write

$$\mathbb{P}(0 \leftrightarrow x) \approx \sum_{n \geq 1} \sum_{\substack{\gamma_1, \dots, \gamma_n \\ D(\gamma_i) = x \\ \text{irred.}}} \mathbb{P}(C_0 = \gamma_1 \amalg \dots \amalg \gamma_n)$$

For $\gamma : y \rightarrow z$, $D(\gamma) = z - y$.

Decomposition into irreducible pieces

We can thus write

$$e^{\xi(x)} \mathbb{P}(0 \leftrightarrow x) \approx \sum_{n \geq 1} \sum_{\substack{\gamma_1, \dots, \gamma_n \\ D(\gamma_i) = x \\ \text{irred.}}} \mathbb{P}(C_0 = \gamma_1 \amalg \dots \amalg \gamma_n) e^{\xi(x)}$$

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 \end{aligned}$$

For $\gamma : y \rightarrow z$, $D(\gamma) = z - y$.

Decomposition into irreducible pieces

Thanks to independence of edge states in Bernoulli percolation,

$$\mathbb{P}(C_0 = \gamma_1 \amalg \dots \amalg \gamma_n) = \prod_{i=1}^n w(\gamma_i),$$

where w is morally given by $w(\gamma) = p^{|\gamma|}(1-p)^{|\partial\gamma|}$.

Decomposition into irreducible pieces

We set, for $x \in \mathbb{Z}^d$,

$$\mathbb{Q}(x) = e^{(t,x)d} \sum_{\substack{\gamma: 0 \rightarrow x \\ \text{irred.}}} w(\gamma).$$

We then have

- \mathbb{Q} is a probability measure on \mathbb{Z}^d ;
- $\mathbb{Q}(|x| > \ell) \leq e^{-c\ell}$, for some $c > 0$.

Decomposition into irreducible pieces

We can thus write

$$e^{\xi(x)} \mathbb{P}(0 \leftrightarrow x) \approx \sum_{n \geq 1} \sum_{\substack{\gamma_1, \dots, \gamma_n \\ D(\gamma_i) = x \\ \text{irred.}}} \mathbb{P}(C_0 = \gamma_1 \amalg \dots \amalg \gamma_n) \prod_{i=1}^n e^{(t, D(\gamma_i))_d}$$

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 &= \sum_{n \geq 1} \sum_{\substack{x_1, \dots, x_n \\ \sum x_i = x}} \prod_{i=1}^n \mathbb{Q}(x_i)
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 &= \sum_{n \geq 1} \sum_{\substack{x_1, \dots, x_n \\ \sum x_i = x}} \prod_{i=1}^n \mathbb{Q}(x_i) \\
 &= \text{Prob}(\exists n \geq 1 : X_n = x),
 \end{aligned}$$

where X is a (directed) random walk on \mathbb{Z}^d with i.i.d. increments of law \mathbb{Q} .

Decomposition into irreducible pieces

Ornstein-Zernike asymptotics now easily follow from the local limit theorem for i.i.d. random variables with small exponential moments:

$$\begin{aligned} e^{\xi(x)} \mathbb{P}(0 \leftrightarrow x) &\approx \text{Prob}(\exists n \geq 1 : X_n = x) \\ &= \frac{C(t)}{|x|^{(d-1)/2}} (1 + o(1)), \end{aligned}$$

where t being dual to x only depends on $x/|x|$.

“Proof by analogy” that \mathbb{Q} is a probability measure on \mathbb{Z}^d :

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$$\implies \mathbb{F}(1) = 1$$

which is equivalent to $\sum_{k \geq 1} f_k = 1$.

More general situations

Some problems with the above argument in more general cases (say, FK percolation with $q > 1$):

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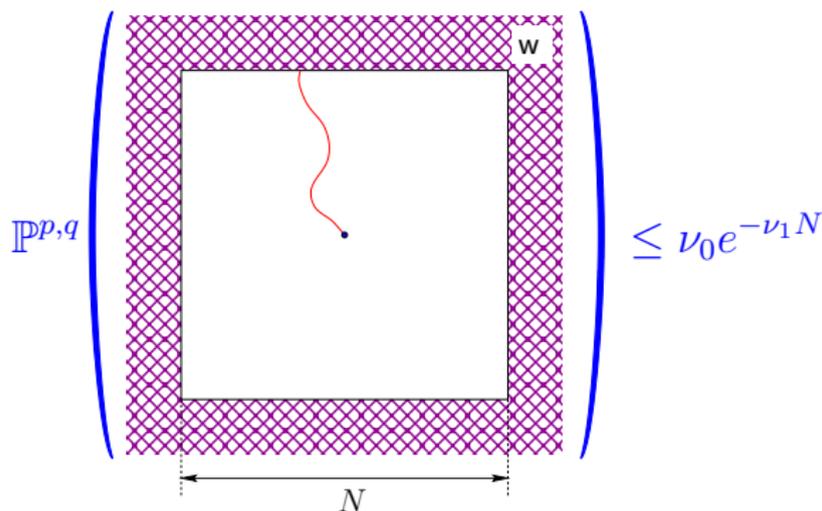
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Both can be dealt with using suitable exponential mixing properties (and extending the local limit theorem from i.i.d. to random variables with exponential mixing).

Absence of BK: FK-percolation with $q > 1$

We *assume* that p is such that there exist $\nu_0, \nu_1 > 0$ s.t., $\forall N$,



Absence of BK: FK-percolation with $q > 1$

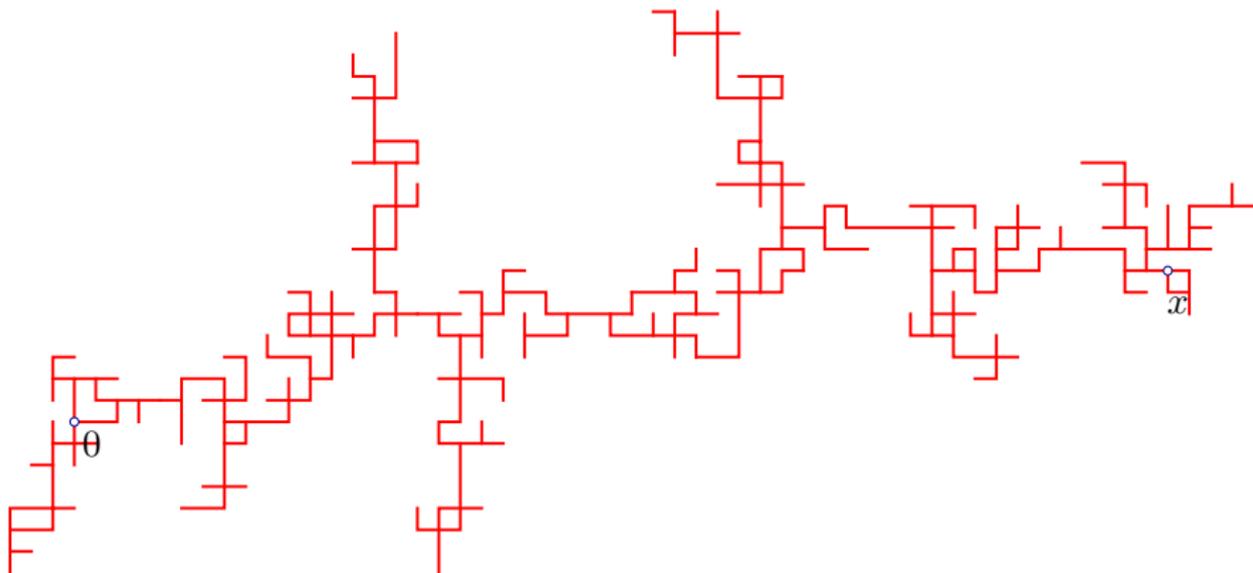
Conjecture

This is true for all $p < p_c(q)$.

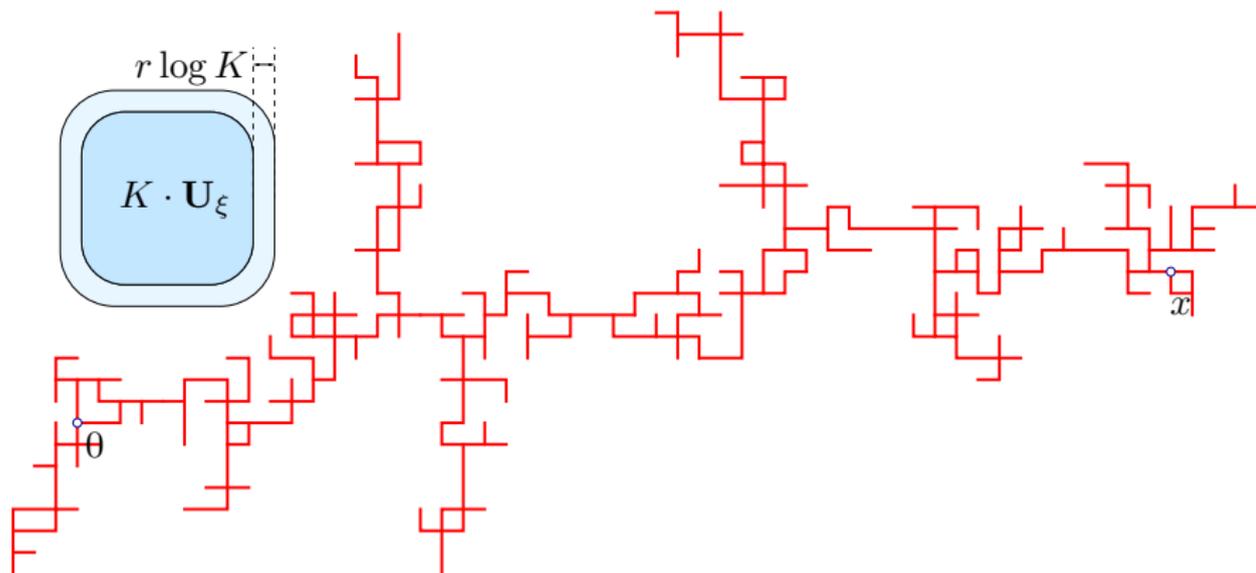
Known ($\forall d$) when:

- $q = 1$ [Aizenman-Barsky '87]
- $q = 2$ [Aizenman *et al* '87]
- $q \gg 1$ [Laanait *et al* '91]

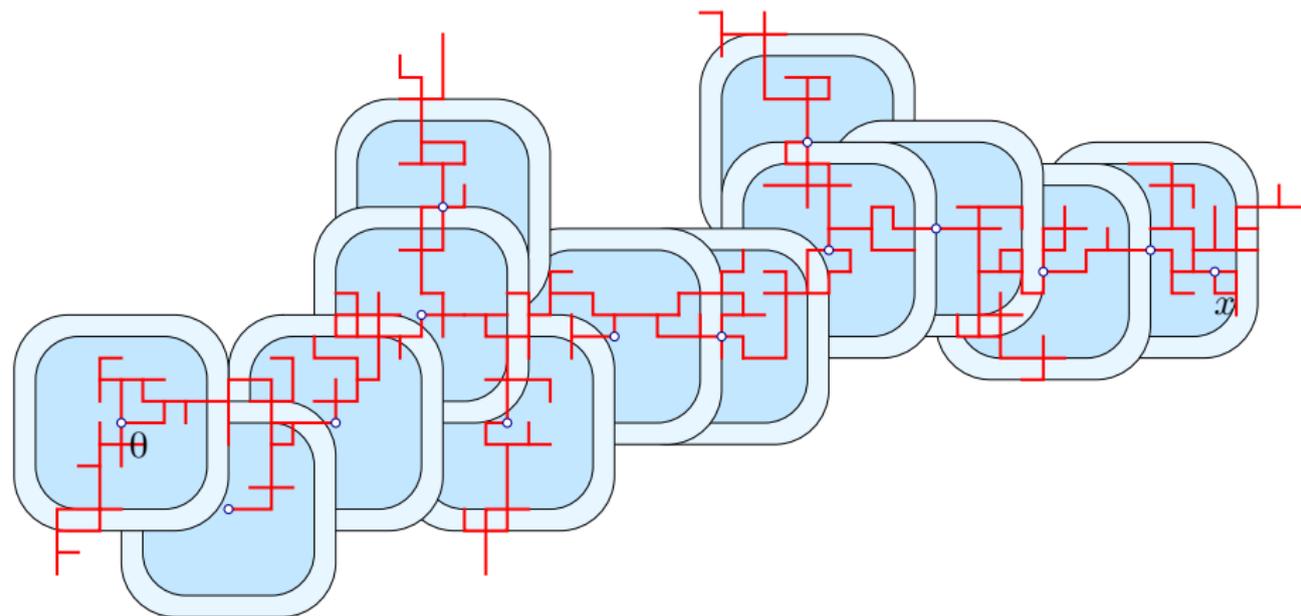
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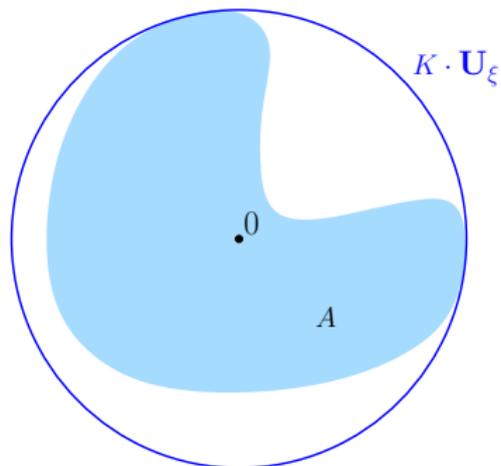
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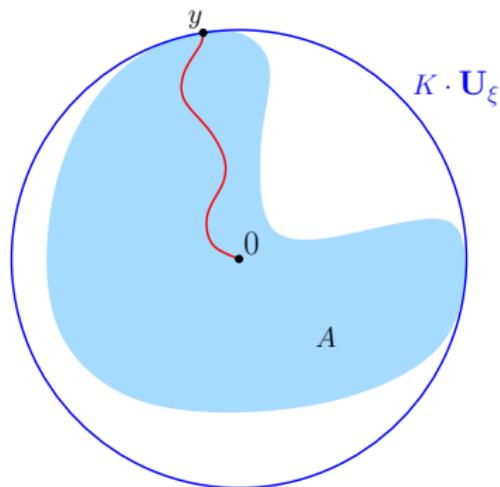
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Mixing for connectivities

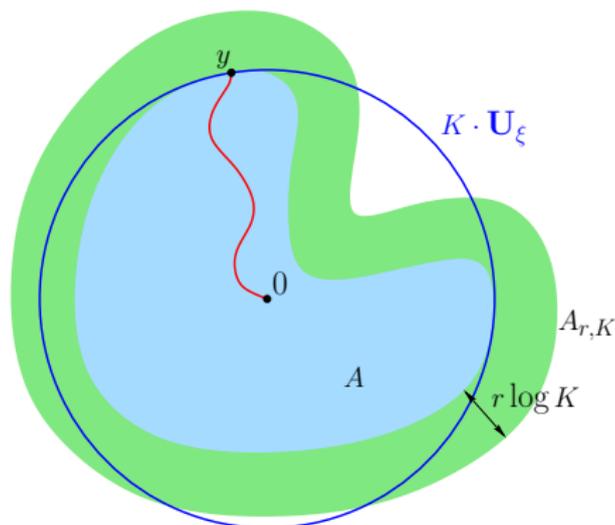


Mixing for connectivities



$$\mathbb{P}^{p,q}(0 \stackrel{A}{\leftrightarrow} y)$$

Mixing for connectivities



$$\sup_{\bar{\omega}} \mathbb{P}^{p,q}(0 \stackrel{A}{\leftrightarrow} y \mid \omega \equiv \bar{\omega} \text{ off } A_{r,K}) \leq e^{-K} (1 + o_K(1))$$

Limit theorem

For $q > 1$, $e^{\xi(x)}\mathbb{P}(C_0 = \gamma_1 \amalg \dots \amalg \gamma_n)$ does not factorize anymore. However, we can write

$$\begin{aligned} e^{\xi(x)}\mathbb{P}(C_0 = \gamma_1 \amalg \dots \amalg \gamma_n) \\ = \mathbb{Q}(\gamma_1)\mathbb{Q}(\gamma_2|\gamma_1) \cdots \mathbb{Q}(\gamma_n|\gamma_1 \amalg \gamma_2 \amalg \dots \amalg \gamma_{n-1}), \end{aligned}$$

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for some suitable measure \mathbb{Q} on finite strings of irreducible paths. Moreover,

$$\frac{\mathbb{Q}(\gamma_k|\gamma_1 \amalg \dots \amalg \gamma_\ell \amalg \gamma_{\ell+1} \amalg \dots \amalg \gamma_{k-1})}{\mathbb{Q}(\gamma_k|\tilde{\gamma}_1 \amalg \dots \amalg \tilde{\gamma}_\ell \amalg \gamma_{\ell+1} \amalg \dots \amalg \gamma_{k-1})} \leq e^{-c(k-\ell)}.$$

Under these conditions, it is possible to extend the local limit theorem.