

UNDERSTANDING LATTICE WALKS VIA CENTRAL WEIGHTINGS

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Séminaire Combinatoire
12 octobre 2016



Co-authors

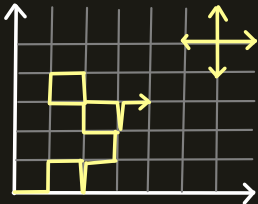
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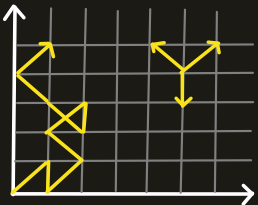
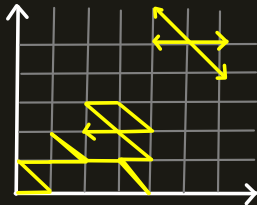
MOTIVATION

$$\sim \frac{4}{\pi} 4^n n^{-1}$$

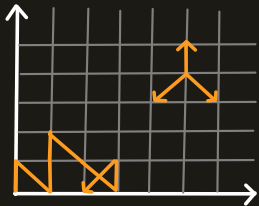


Various asymptotic behaviours

$$\sim \frac{8}{\pi} 4^n n^{-2}$$



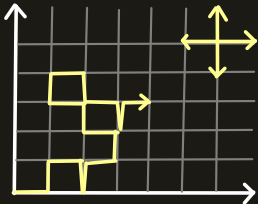
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$$\sim \frac{4A_n}{\pi} (2\sqrt{2})^n n^{-2}$$

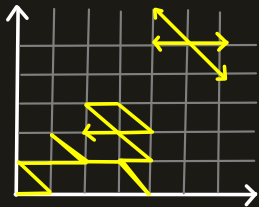
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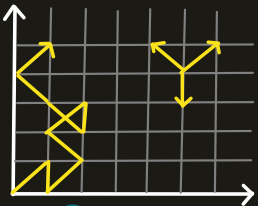


Various asymptotic behaviours
Exponential growth

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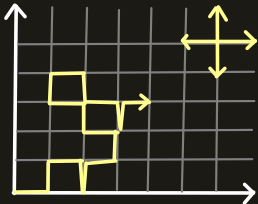
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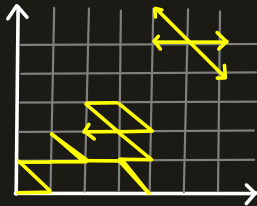
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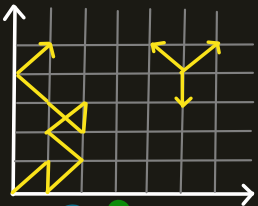


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Exponential growth
Critical exponent

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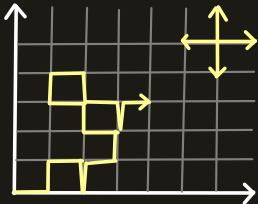
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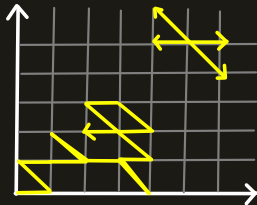
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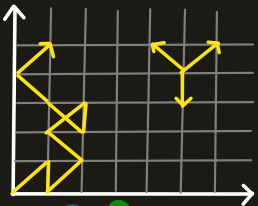
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Motivation: opt for a continuous model to detect the transitions

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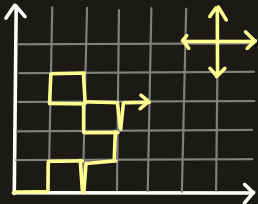


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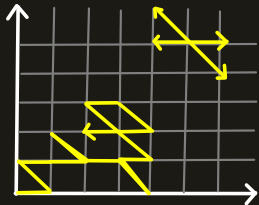
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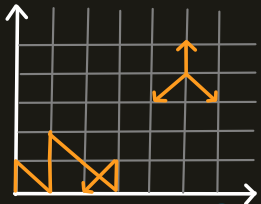
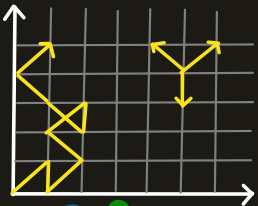
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Motivation: opt for a continuous model to detect the transitions

Central weightings

$$\sim \frac{\sqrt{3}}{2\sqrt{\pi}} 3^n n^{-\frac{1}{2}}$$



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PART 1

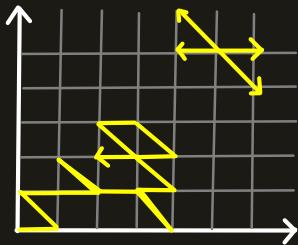
Weighted Gouyou-Beauchamps model

GOUYOU-BEAUCHAMPS MODEL

In general: walks starting at $(0,0)$
staying in the positive quadrant
with steps in \mathcal{F}

where $\mathcal{F} \subseteq \{\leftarrow, \rightarrow, \uparrow, \downarrow, \nearrow, \searrow, \swarrow, \nwarrow\}$

Here: $\mathcal{F} = \{\leftarrow, \rightarrow, \nearrow, \searrow\}$
→ Gouyou-Beauchamps model



Number of walks ending anywhere after...

1 step: 1 2 steps: 3 3 steps: 6
4 steps: 20 5 steps: 50 6 steps: 175

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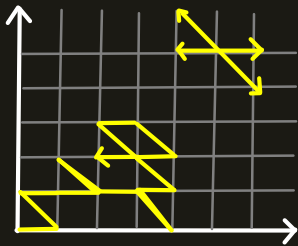
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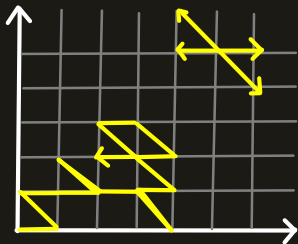
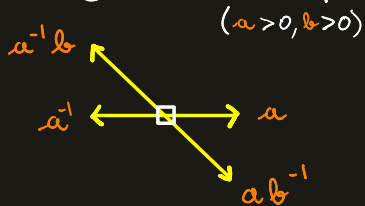
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4 steps: 20 5 steps: 50 6 steps: 175



A small connection
with probabilities:
 $\mathbb{P} \left(\begin{array}{c} \text{staying in the} \\ \text{quadrant after} \\ \text{6 steps} \end{array} \right) = \frac{175}{4^6}$

WEIGHTED GOUYOU-BEAUCHAMPS MODEL

A weight to each step:



Weight of walks ending anywhere after...

1 step: a

2 steps: $1 + b + a^2$

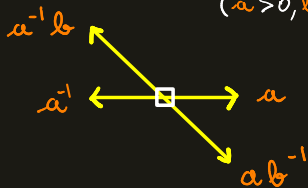
3 steps: $2ab + a^3 + 3a$

A small connection
with probabilities:

$$\mathbb{P} \left(\begin{array}{l} \text{staying in the} \\ \text{quadrant after} \\ \text{3 steps} \end{array} \right) = \frac{2ab + a^3 + 3a}{(a + a^{-1} + ab^{-1} + ba^{-1})^3}$$

WEIGHTED GOUYOU-BEAUCHAMPS MODEL

A weight to each step:
 $(a > 0, b > 0)$



Weight of walks ending anywhere after..

$$1 \text{ step: } a = [z_0^1] Q_{\uparrow}(1, 1; z)$$

$$2 \text{ steps: } 1 + b + a^2 = [z_0^2] Q_{\uparrow}(1, 1; z)$$

$$3 \text{ steps: } 2ab + a^3 + 3a = [z_0^3] Q_{\uparrow}(1, 1; z)$$

$Q_{\uparrow}(x, y; z) :=$ generating function
of weighted GB-walks

$$= \sum_{\substack{\text{w GB-walk} \\ (0,0) \rightarrow (i,j) \\ \text{of length } n}} \text{weight}(w) x^i y^j z^n$$

$$\text{weight}(w) := \prod_{\substack{\Delta \text{ step in } w \\ \text{(with multiplicity)}}} \text{weight}(\Delta)$$

$Q(x, y; z) =$ GF of unweighted GB-walks

$$= \sum_{\substack{\text{w GB-walk} \\ (0,0) \rightarrow (i,j) \\ \text{of length } n}} x^i y^j z^n$$

Rk: $Q_{\uparrow}(1, 1; z) = Q(a, b; z)$

CENTRAL WEIGHTING

Theorem A weight μ_Δ is assigned to each $\Delta \in \{\leftarrow, \rightarrow, \uparrow, \downarrow\}$.
There is equivalence between:

(i) The probability of a given walk only depends on its length and its endpoint.

(ii) There exist constants $a, b, c > 0$ such that
 $\mu_{\leftarrow} = c a^{-1}$ $\mu_{\rightarrow} = c a$ $\mu_{\uparrow} = c a^{-1} b$ $\mu_{\downarrow} = c a b^{-1}$

(iii) $\mu_{\leftarrow} \times \mu_{\rightarrow} = \mu_{\uparrow} \times \mu_{\downarrow}$

(iv) If $Q_{\uparrow}(\alpha, y; z) :=$ the weighted generating function
and $Q(\alpha, y; z) :=$ the unweighted one,
then $Q_{\uparrow}(\alpha, y; z) = Q(a\alpha, by; cz)$.

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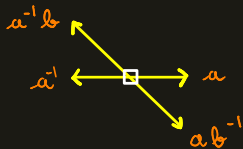
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Too  restrictive?

THE BIG THEOREM

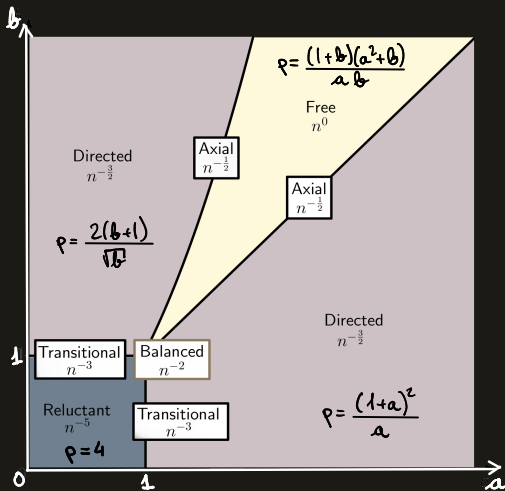


Theorem :

Weight of GB-walks of length n

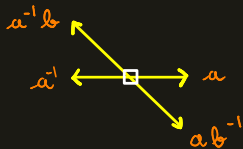
$$= [z^n] Q_{\uparrow}(1, 1; t)$$

$$\sim K^{[n]} \rho^n n^{-\alpha}$$



THE BIG THEOREM

In terms of the drift = $(a - a^{-1} + ab^{-1} - a^{-1}b, a^{-1}b - ab^{-1})$

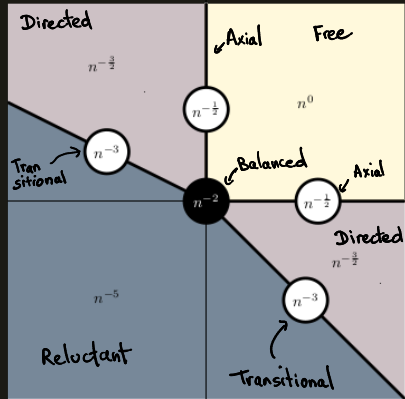


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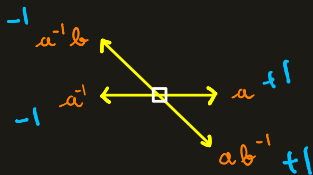
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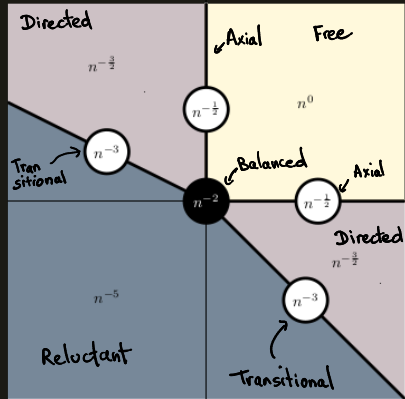


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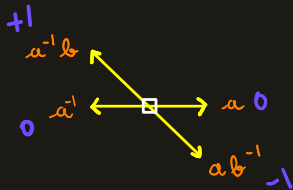
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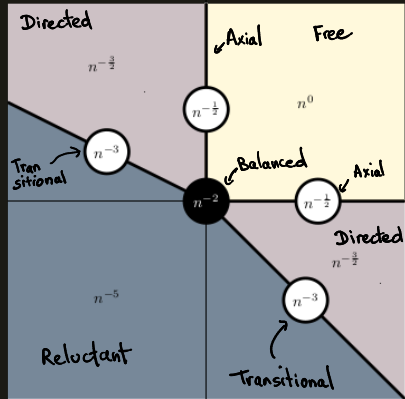
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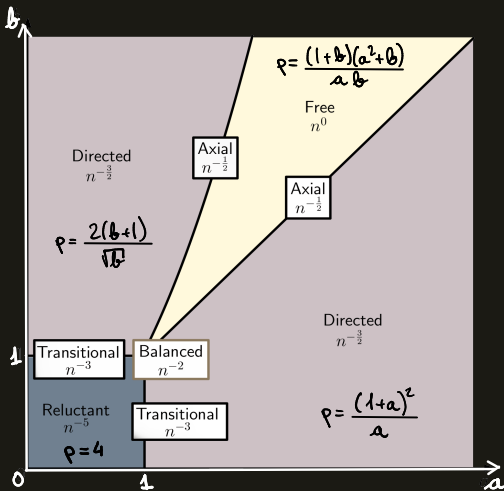
Also works when $(0,0)$ is not the starting point.

Theorem :

Weight of GB-walks of length n starting at (i,j) ending anywhere

$$\sim K \sqrt{\binom{n}{i,j}} \rho^n n^{-\alpha}$$

↑
fonction harmonique



APPLICATION : RANDOM GENERATION

① [Lumbroso, Mishna, Ponty]

Generation in $O(n^{\alpha-\frac{1}{2}} \ln n)$
time



generation in a
half-plane
⊕ rejection if
not the quadrant


② Anticipated rejection
algorithm from
[Bacher, Sportiello]



Linear complexity
for free and axial
cases

UNDERSTANDING THE PROOF

Main ingredient :

in  Analytic
Combinatorics
Several
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Steps:

- ① Express the generating function in terms of a diagonal
- ② Find the contributing points ← exponential growth
- ③ Transform the generating function into an integral and apply some theorem ([Hörmander]) ← whole asymptotic estimate

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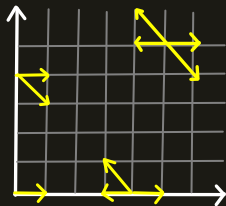
FUNCTIONAL EQUATION

3 important functions:

$$S(x, y) = x + x^{-1} + xy^{-1} + x^{-1}y \quad \text{INVENTORY}$$

$$K(x, y; z) = xy(1 - zS(x, y)) \quad \text{KERNEL}$$

$$Q(x, y; z) \quad \text{UNWEIGHTED GENERATING FUNCTION}$$



KERNEL EQUATION:

[Bousquet-Mélou, Mishna]

$$K(x, y; z) Q(x, y; z) = xy - K(x, 0; z) Q(x, 0; z) - K(0, y; z) Q(0, y; z) + K(0, 0; z) Q(0, 0; z)$$

Walks without constraints

Boundary restrictions
(Inclusion-exclusion)

INTO A DIAGONAL EXPRESSION

KERNEL EQUATION:

$$K(x, y; z) Q(x, y; z) = xy - \underbrace{K(x, 0; z) Q(x, 0; z) - K(0, y; z) Q(0, y; z) + K(0, 0; z) Q(0, 0; z)}$$

We want to eliminate this

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Algebraic operations
 (Orbit sum
 ⊕ Extraction of
 positive coefficients)

DIAGONAL EXPRESSION:

$$Q(1, 1; z) = \Delta \left(\frac{(1-x)(1-y)(1+x)(x^{-2}-y^{-1})(x-y)(x+y)}{(1-xy) S(x^{-1}, y^{-1}) (1-x)(1-y)} \right)$$

where $\Delta \left(\sum_{i, j, n \geq 0} f_{i, j, n} x^i y^j z^n \right) = \sum_{n \geq 0} f_{n, n, n} z^n$

INTO A DIAGONAL EXPRESSION

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orbit sum
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codes the excursions

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INTO A DIAGONAL EXPRESSION

KERNEL EQUATION:

$$K(x, y; z) Q(x, y; z) = xy - \underbrace{K(x, 0; z) Q(x, 0; z) - K(0, y; z) Q(0, y; z) + K(0, 0; z) Q(0, 0; z)}$$

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orbit sum applied to the start point

codes the excursions

releases the x -constraint

releases the y -constraint

where

$$\Delta \left(\sum_{i, j, n \geq 0} f_{i, j, n} x^i y^j z^n \right) = \sum_{n \geq 0} f_{n, n, n} z^n$$

WEIGHTED VERSION

$$Q_{\uparrow}(1,1; \frac{a}{b}, \frac{x}{y}) = \frac{1}{a^4 b^3 \frac{x^2}{y^2}} \Delta \left(\frac{y^2 (y-b)(a-x)(a+x)(a^2 y - b x^2)(a y - b x)(a y + b x)}{(1 - x y \frac{x}{y} S(x^{-1}, y^{-1})) (1-x) (1-y)} \right)$$

where

$$S(x, y) = a x + a^{-1} x^{-1} + a b^{-1} x y^{-1} + a^{-1} b x^{-1} y \quad \text{WEIGHTED INVENTORY}$$

WEIGHTED VERSION


$$Q_{\uparrow}(1,1;g) = \frac{1}{a^4 b^3 z^2} \Delta \left(\underbrace{y^2 z \frac{(y-b)(a-x)(a+x)(a^2 y - b x^2)(ay - bx)(ay + bx)}{(1 - axyz S(x^{-1}, y^{-1})) (1-x) (1-y)}}_{:= F(x,y,g) = \frac{G(x,y,g)}{H(x,y,g)}} \right)$$

where

$$S(x,y) = ax + a^{-1}x^{-1} + ab^{-1}xy^{-1} + a^{-1}b^{-1}x^{-1}y^{-1} \quad \text{WEIGHTED INVENTORY}$$

UNDERSTANDING THE PROOF

Main ingredient :

in  Analytic
Combinatorics
Several
Variables

Steps:

- ① Express the generating function ✓
in terms of a diagonal
- ② Find the contributing points ← exponential growth
- ③ Transform the generating function
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blackboard

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AN EXPRESSION FOR THE RADIUS

Recall: $Q_n(1,1; z_0) = \Delta F(x,y,z_0) = \Delta \frac{G(x,y,z_0)}{H(x,y,z_0)}$

Quick reasoning:

Let \mathcal{D} = disk of convergence of F , and $(x,y,z_0) \in \mathcal{D}$.

F is absolutely convergent on (x,y,z_0)

$$\text{so } Q_n(1,1; |x,y,z_0|) < +\infty$$

Hence radius of $Q_n \geq \sup_{x \in \overline{\mathcal{D}}} |x,y,z_0|$

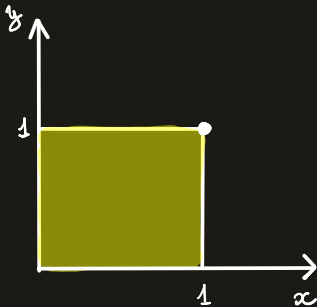
Actually $\text{radius of } Q_n = \sup_{\substack{x \in \overline{\mathcal{D}} \\ H(x,y,z_0) \neq 0}} |x,y,z_0|$

AN EXPRESSION FOR THE RADIUS

I should have convinced you that:

$$\left(\text{radius of } Q_p(1,1; \mathbb{R}^2)\right)^{-1} = \min_{\substack{|x| \leq 1 \\ |y| \leq 1}} |S(x^{-1}, y^{-1})|$$

$S(x, y) = \text{WEIGHTED INVENTORY}$

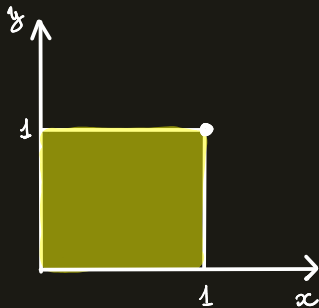


AN EXPRESSION FOR THE RADIUS

I should have convinced you that:

$$\left(\text{radius of } Q_p(1,1; \sigma)\right)^{-1} = \min_{\substack{|x| \leq 1 \\ |y| \leq 1}} |S(x^{-1}, y^{-1})|$$

$S(x, y) = \text{WEIGHTED INVENTORY}$



If minimum \in ,

$a < 1$ and $b < 1$

reluctant case

If minimum \in  or ,

$a \geq 1$ and $a > b$
or $b \geq 1$ and $b > a^2$.

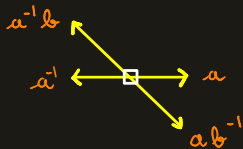
directed case

If minimum \in ,

$1 < \sqrt{b} < a < b$

Free case

THE BIG THEOREM

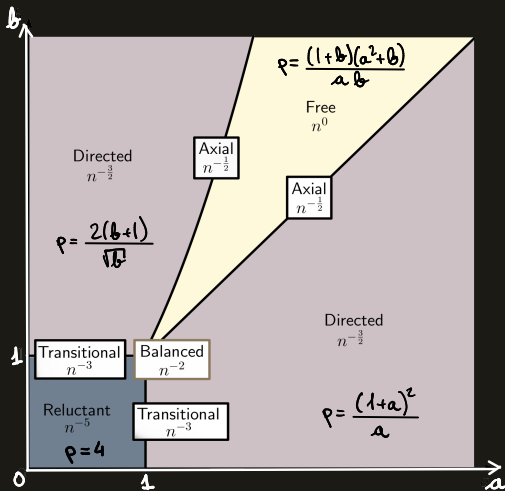


Theorem :

Weight of GB-walks of length n

$$= [z^n] Q_{\uparrow}(1, 1; t)$$

$$\sim K^{[n]} \rho^n n^{-\alpha}$$



PART 2

General Central Weightings

CENTRAL WEIGHTINGS IN ALL GENERALITY

Theorem Let \mathcal{S} be a non-singular set of integer steps in dimension 2.

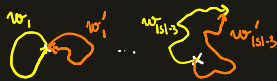
A weight μ_Δ is assigned to each $\Delta \in \mathcal{S}$. There is equivalence between

(i) The probability of a given walk only depends on its length, its start and end point.

(ii) There exist constants $a, b, c > 0$ such that
for every $(x_\Delta, y_\Delta) \in \mathcal{S}$, $\mu_{(x_\Delta, y_\Delta)} = c \times a^{x_\Delta} \times b^{y_\Delta}$

(iii) Take $|\mathcal{S}| - 2 - 1$ "independent" pairs of paths (w_k, w'_k) such that w_k and w'_k share the same length, the same start and end point.

Then for every k ,

$$\prod_{\substack{\Delta \in w_k \\ \text{with multiplicity}}} \mu_\Delta = \prod_{\substack{\Delta' \in w'_k \\ \text{with multiplicity}}} \mu_{\Delta'}$$


(iv) The kernels are essentially the same.

GENERATING FUNCTIONS

What about

(10) There exist constants $a, b, c > 0$ such that

$$Q_p(x, y; z) = Q(ax, by; cz)$$

↑ weighted GF ↑ unweighted GF

?

GENERATING FUNCTIONS

What about

(10) There exist constants $a, b, c > 0$ such that

$$Q_p(x, y; z) = Q(ax, by; cz)$$

↑ weighted GF ↑ unweighted GF

Prop: If a weighting is central, then (10) holds.

Conjecture: If (10) holds, then the weighting is central.

(true for all small 2D-models)

SOME CONSEQUENCES

(10) There exist constants $a, b, c > 0$ such that

$$Q_{\mu}(x, y; z) = Q(ax, by, cz)$$

↑ weighted GF ↑ unweighted GF

Consequence 1:

$Q_{\mu}(x, y; z)$ is \mathbb{D} -finite.

\Leftrightarrow

$Q(x, y; z)$ is \mathbb{D} -finite.

(if the weights are central
& rational)

SOME CONSEQUENCES

(10) There exist constants $a, b, c > 0$ such that

$$Q_{\mu}(x, y; z) = Q(ax, by, cz)$$

↑ weighted GF
↑ unweighted GF

Consequence 1: $Q_{\mu}(x, y; z)$ is \mathbb{D} -finite.

\iff
 $Q(x, y; z)$ is \mathbb{D} -finite.

(if the weights are central
 \oplus rational)

Consistant with [Kauers, Yatchak]: systematic search of the \mathbb{D} -finite weighted 2D-models

Family 1a



Family 1b



Family 2a



Family 2b



Family 3a



Family 3b



Family 4a



Family 4b



SOME CONSEQUENCES

(10) There exist constants $a, b, c > 0$ such that

$$Q_{\mu}(x, y; z) = Q(ax, by, cz)$$

↑ weighted GF ↑ unweighted GF

Consequence 2: Given a central weighting $\mu(x_s, y_s) = c \cdot a^{x_s} b^{y_s}$,

Weight of excursions of size $n = c^n \times$ number of excursions of size n .

SOME CONSEQUENCES

(15) There exist constants $a, b, c > 0$ such that

$$Q_{\mu}(x, y; z) = Q(ax, by, cz)$$

↑ weighted GF ↑ unweighted GF

Consequence 2: Given a central weighting $\mu(x_s, y_s) = c \cdot a^{x_s} b^{y_s}$,
Weight of excursions of size $n = c^n \times$ number of excursions of size n .

In general,



THE CONJECTURE OF GARBIT, MUSTAPHA & RASCHEL

$S(x, y)$ = WEIGHTED INVENTORY

$$(x^*, y^*) = \underset{\substack{x \geq 1 \\ y \geq 1}}{\operatorname{argmin}} S(x, y)$$

Conjecture (Every estimate is up to a constant.)

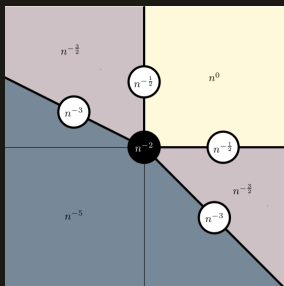
	$\nabla S(x^*, y^*) = 0$	or $\frac{\partial S}{\partial x}(x^*, y^*) > 0$ $\frac{\partial S}{\partial y}(x^*, y^*) > 0$	and $\frac{\partial S}{\partial x}(x^*, y^*) > 0$ $\frac{\partial S}{\partial y}(x^*, y^*) > 0$
$(x^*, y^*) = (1, 1)$	$\sim S(1, 1)^n \sim n^{-p/2}$ balanced	$\sim S(1, 1)^n \sim n^{-1/2}$ axial	$\sim S(1, 1)^n \sim n^0$ free
$x^* = 1$ or $y^* = 1$	$\sim S(x^*, y^*)^n \sim n^{-p/2+1}$ transitional	$\sim S(x^*, y^*)^n \sim n^{-3/2}$ directed	impossible
$x^* > 1$ and $y^* > 1$	$\sim S(x^*, y^*)^n \sim n^{-p-1}$ reluctant	impossible	impossible

$$\mu = \pi / \arccos(-c)$$

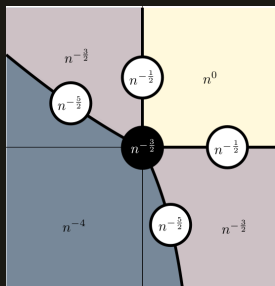
$$c = \frac{\partial^2 S}{\partial x \partial y}(x^*, y^*) / \sqrt{\frac{\partial^2 S}{\partial x^2}(x^*, y^*) + \frac{\partial^2 S}{\partial y^2}(x^*, y^*)}$$

do not depend on the central weights

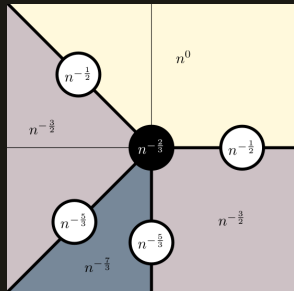
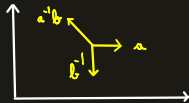
SOME DRIFT DIAGRAMS



Gouyou - Beauchamps model



Tandem model



Gessel model in its natural cone.



CONCLUSION

→ Good framework to understand transitions -

→ Many left conjectures:

- A weighting is central iff $Q_{\mu}(x, y; z) = Q(ax, by; cz)$?

- Which step sets have "pretty" (like conic) regions in their drift diagrams?

- ...

→ Hope to understand more non-D-finite models.



The image shows a screenshot of a Scratch-like programming environment. On the left, a stage displays a drawing of a jagged path in blue and green, with a cat sprite at the top. Below the stage, a palette shows two sprites: 'Lutin1' (a cat) and 'Bat1' (a bat). The right side of the screen shows the code editor with the following script:

```
quand cliqué  
effacer tout  
répéter 100 fois  
  aller à x: -240 y: -180  
  stylo en position d'écriture  
  ajouter 10 à couleur du stylo  
  répéter jusqu'à 180 < ordonnée y ou 240 < abscisse x ou ordonnée y < -180  
    mettre alca à nombre aléatoire entre 1 et 6  
    si alca = 1 ou alca = 2 alors  
      avancer de 10  
    si alca = 3 alors  
      avancer de -10  
    si alca = 4 alors  
      tourner de 45 degrés  
      avancer de 10  
      s'orienter à 90°  
    si alca = 5 ou alca = 6 alors  
      tourner de 45 degrés  
      avancer de -10  
      s'orienter à 90°  
relevé le stylo
```