

*KPZ wandering exponent for random walk  
in i.i.d. dynamic Beta random environment*

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*Joint work with Márton Balázs (Bristol)  
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## Coin tosses and random walk

Toss a coin: Heads with probability  $p$ , Tails with probability  $1 - p$ .

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Repeated tosses  $\longleftrightarrow$  up-right path (Random Walk on  $\mathbb{Z}^2$ ):

HHTHTTTTHHTHT  $\longleftrightarrow$



## Classical results: LLN, CLT, LDP

$X_0 = 0$ ,  $X_n$  = position on up-right path after  $n$  tosses/steps.

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Large Deviation Principle (LDP):  $P(\text{proportion of } H \geq s > p) \approx e^{-nH(s)}$

$-n^{-1} \log P\{X_n \cdot e_1 \geq sn\} \rightarrow H(s) = s \log \frac{s}{p} + (1-s) \log \frac{1-s}{1-p}$

$H(s) =$  entropy of coin  $s$  relative to coin  $p$ .

## Conditioned random walk

Large deviations also tell us that  $X_{0,n} = (X_0, \dots, X_n)$  conditioned on

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New random walk still has CLT fluctuations (of size  $\sqrt{n}$ ).

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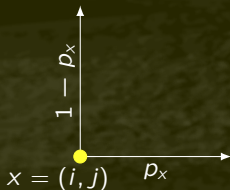
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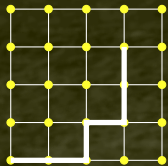
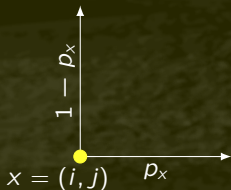
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*HHTHTT* has probability

$$p_{0,0}p_{1,0}(1 - p_{2,0})p_{2,1}(1 - p_{3,1})(1 - p_{3,2})$$

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Also, Quenched CLT (R-A, Seppäläinen '05): for almost every environment  $\{p_x : x \in \mathbb{Z}_+^2\}$

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Note: once environment is fixed,  $X_n$  is no longer a random walk with i.i.d. increments.

## LLN, CLT, LDP

Averaged LDP: when environment is averaged out and  $s > \bar{p}$

$$-n^{-1} \log P\{X_n \cdot e_1 \geq sn\} \rightarrow H_a(s) = s \log \frac{s}{\bar{p}} + (1 - s) \log \frac{1 - s}{1 - \bar{p}}.$$

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$H_q$  is deterministic but in general does not have an explicit expression (though some variational formulas are available).

$H_q(s) > H_a(s)$  unless  $s = \bar{p}$ , in which case both = 0.



## Solvable model

Explicit computations are possible when  $p_x \sim \text{Beta}(\alpha, \beta)$ ,  $\alpha, \beta > 0$ .

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Can also compute the quenched rate  $H_q(s)$  explicitly (later).

## KPZ fluctuation exponent

*Barraquand and Corwin '15 observed a connection to KPZ:*

**Theorem.** *For the Beta( $\alpha, \beta$ ) case*

$$\frac{\log P^\omega\{X_n \cdot e_1 \geq sn\} + nH_q(s)}{\sigma(s)n^{1/3}} \longrightarrow GUE \quad (\text{in distribution})$$

*( $\sigma(s)$  is known explicitly in terms of polygamma functions  $\psi_1$  and  $\psi_2$ ).*

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*But how could it? We know the CLT holds, both quenched and averaged!*

*What is going on?!*

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Averaged: average out the environment in the above.

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## Busemann function

**Theorem** (Balázs, R-A, Seppäläinen '16). For almost every choice of the environment  $\omega = \{p_x : x \in \mathbb{Z}_+\}$ , limit

$$B^\zeta(x, y) = \lim_{n \rightarrow \infty} \left[ \log P^\omega(X_n \approx n\zeta \mid X_0 = x) - \log P^\omega(X_n \approx n\zeta \mid X_0 = y) \right]$$

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This comes from the Markov property

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(then divide by  $P^\omega(X_n \approx n\zeta \mid X_0 = 0)$  and take  $n \rightarrow \infty$ ).

## Quenched conditioned RWRE

Define  $\pi^\zeta$  as a Doob transform of  $p$  by the harmonic function  $e^{-B^\zeta(0,x)}$ :

$$\pi_{x,x+e_1}^\zeta = p_x \frac{e^{-B^\zeta(0,x+e_1)}}{e^{-B^\zeta(0,x)}} \quad \text{and} \quad \pi_{x,x+e_2}^\zeta = (1 - p_x) \frac{e^{-B^\zeta(0,x+e_2)}}{e^{-B^\zeta(0,x)}}.$$

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So, if  $\zeta \neq \xi$ , the new process is another random walk in a stationary but very correlated random environment.

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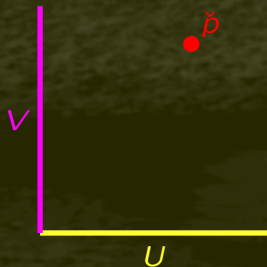
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Let  $\{U_{ke_1} : k \geq 0\}$  be i.i.d.  $\text{Beta}(\alpha + \lambda, \beta)$ .

Let  $\{V_{ke_2}^{-1} : k \geq 0\}$  be i.i.d.  $\text{Beta}(\lambda, \alpha)$ .

Let  $\{\check{\rho}_x : x \in \mathbb{N}^2\}$  be i.i.d.  $\text{Beta}(\alpha, \beta)$ .

All three families are mutually independent.

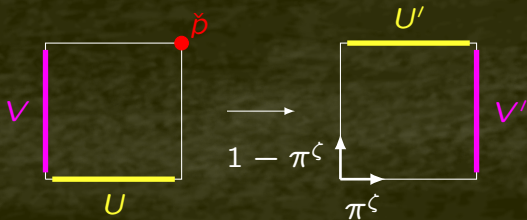


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And define  $\pi_{x, x+e_1}^\zeta = \frac{V_x - 1}{V_x - U_x} \in (0, 1)$  and  $\pi_{x, x+e_2}^\zeta = 1 - \pi_{x, x+e_1}^\zeta$ .

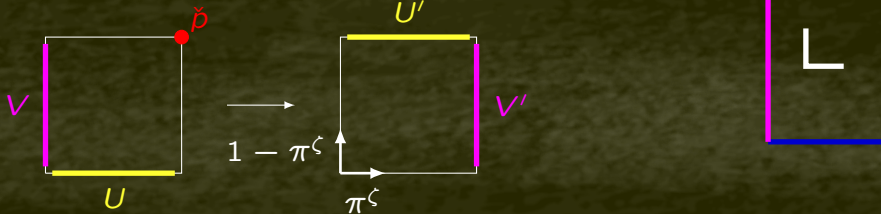


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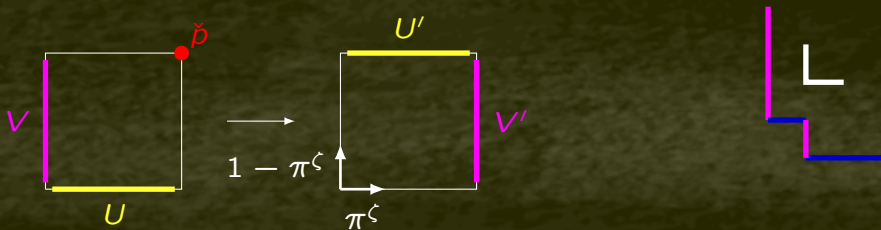


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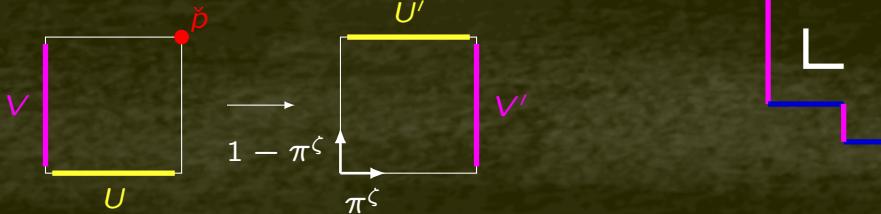


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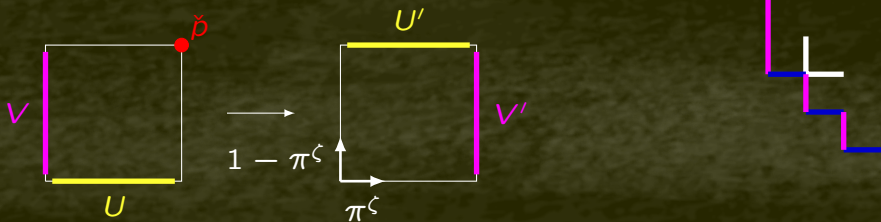


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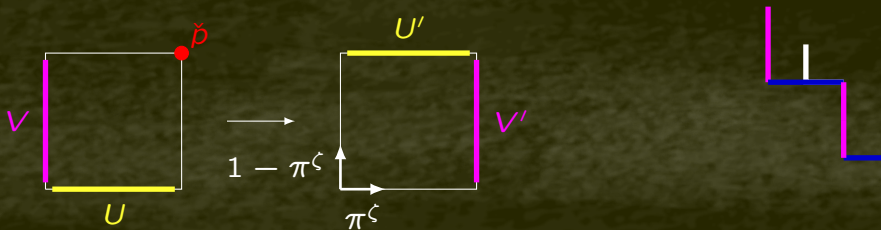


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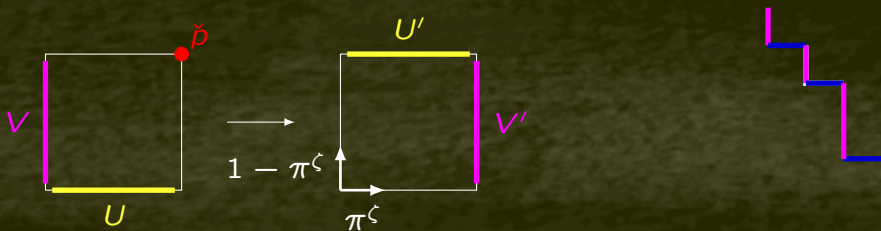


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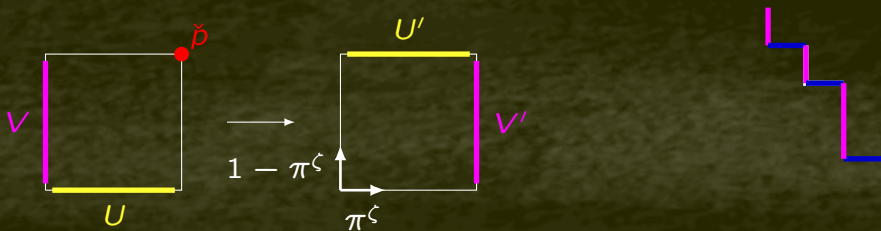


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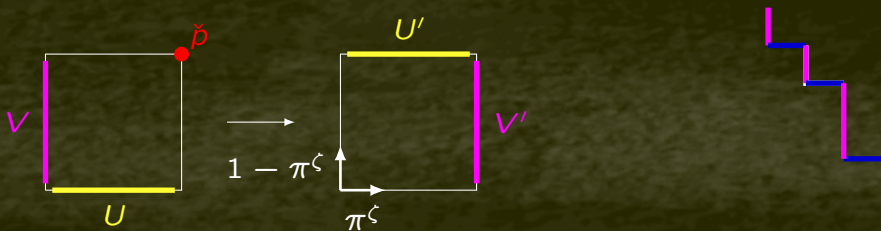
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**Lemma:**  $(U', V', \pi^\zeta)$  has the same distribution as  $(U, V, \check{p})$ .

**Corollary:**  $\{\pi_{x+y, x+y+e_1}^\zeta : y \in \mathbb{Z}_+^2\}$  has the same distribution as for  $x = 0$ .

## Bijection between velocity $\zeta$ and boundary parameter $\lambda$

$\lambda \in [0, \infty]$  is in one-to-one correspondence with  $\zeta$  via

$$\zeta_1 = \frac{\psi_1(\lambda) - \psi_1(\alpha + \lambda)}{\psi_1(\lambda) - \psi_1(\alpha + \beta + \lambda)} \in \left[ \frac{\alpha}{\alpha + \beta}, 1 \right], \quad \zeta_2 = 1 - \zeta_1$$

with  $\lambda = 0 \iff \zeta = e_1$  and  $\lambda = \infty \iff \zeta = \xi = \left( \frac{\alpha}{\alpha + \beta}, \frac{\beta}{\alpha + \beta} \right)$ .

$\psi_1$  is the trigamma function:  $\psi_1 = (\log \Gamma)''$ .

For rest of velocities,  $\zeta_1 \in [0, \frac{\alpha}{\alpha + \beta}]$ , switch role of  $U$ s and  $V$ s.

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$(B^\zeta(0, e_1), B^\zeta(0, e_2)) \sim (\log U_0, \log V_0)$  with parameter  $\lambda(\zeta)$ .



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for  $s \in [\frac{\alpha}{\alpha+\beta}, 1]$ .

(For  $s \in [0, \frac{\alpha}{\alpha+\beta})$  switch the role of the axes.)

(Barraquand and Corwin '15 got this formula first, by a more direct computation.)

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**Theorem:** For  $\zeta \neq \xi$ ,  $\exists C, c: \forall n \in \mathbb{N}$  and  $b$  large,

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Complements the aforementioned results saying KPZ fluctuations exponent (for  $\log P^\omega(X_n \approx n\zeta)$ ) is  $1/3$ .



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*In both models, solvability comes from the Beta-Gamma algebra. Namely:*

*If  $A$  is  $\text{Gamma}(a + b, c)$  and  $B$  is an independent  $\text{Beta}(a, b)$ , then  $AB$  and  $A(1 - B)$  are independent  $\text{Gamma}(a, c)$  and  $\text{Gamma}(b, c)$ .*

## Existence of Busemann limit: coupling

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So we can use these as transitions for the Beta RWRE.



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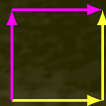
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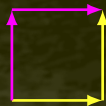
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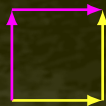
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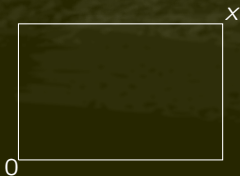


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Then define  $B^\zeta(x, y) = B^\zeta(0, y) - B^\zeta(0, x)$  and we have the cocycle property:  $B^\zeta(x, y) + B^\zeta(y, z) = B^\zeta(x, z)$ .

## Existence of Busemann limit: dual polymer

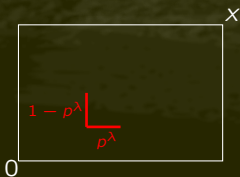
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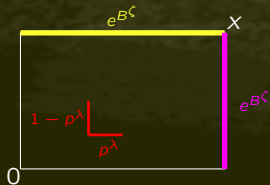
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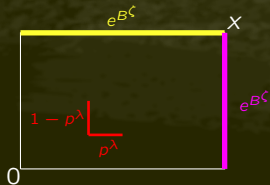


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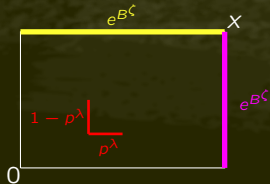


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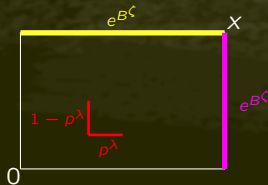
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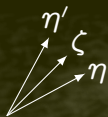
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**Remark:** This connects the RWRE to a polymer with boundary conditions, which leads to the KPZ wandering exponent.

## Existence of Busemann limit: comparison Lemma

By a monotonicity of  $B^\zeta(0, x)$  in the edge weights  $\sigma$  the above gives:

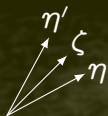
**Lemma:** With probability one, for  $n$  large and  $\eta' \cdot e_1 < \zeta \cdot e_1 < \eta \cdot e_1$   
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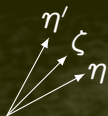
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Thank You!