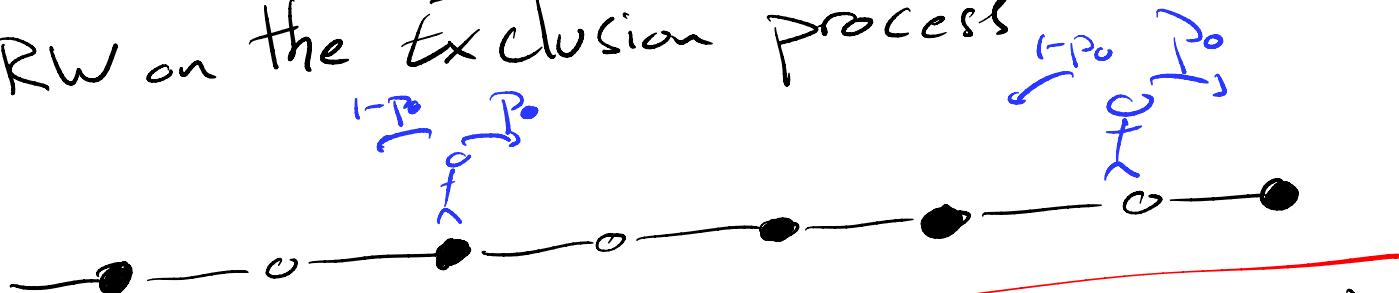


## Lecture 8

RW on the Exclusion process



$(\eta_t)_{t \geq 0}$  Exclusion with  $P^e$

density  $\rho \in (0, 1)$

rate  $v > 0$

$(X_n)_{n \geq 0} : P^L$ : quenched law of  $(X_n)$

$$P^L = E^S \times P^e(\cdot),$$

$$\{0 < P^o \leq P^e < 1\}$$

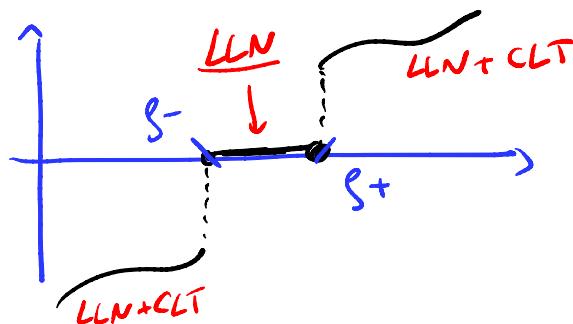
Th:  $\forall s \in (0,1) \setminus \{s^-, s^+\}$

$$\frac{X_n}{n} \rightarrow v(s) \quad \text{P}^s\text{-a.s.}$$

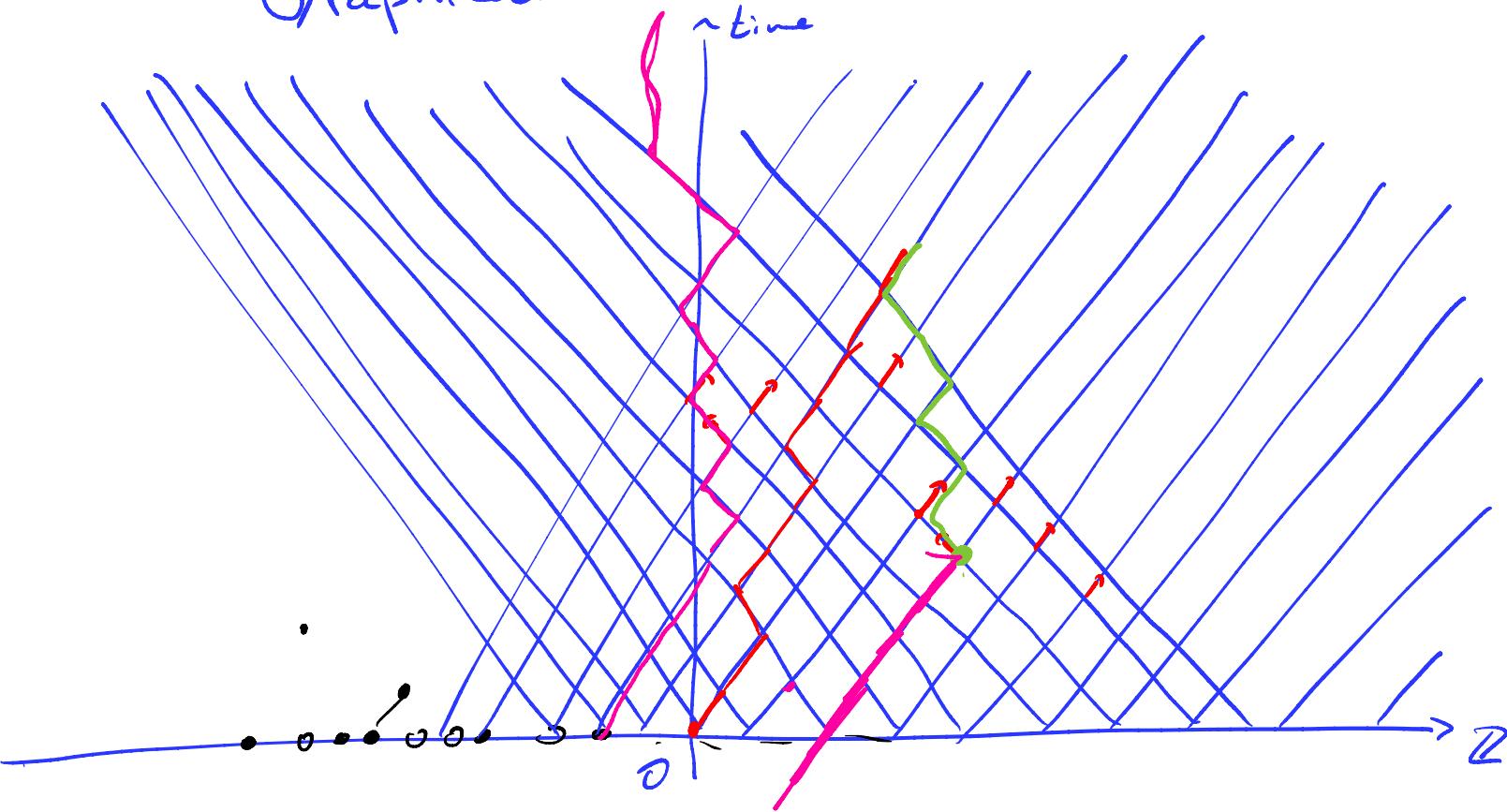
+ CLT for  $s \in [0,1] \setminus [s^-, s^+]$

$$s^- = \sup \{ s : v(s) < 0 \}$$

$$s^+ = \inf \{ s : v(s) > 0 \}$$



Let's "prove" the LCN -  
Graphical construction / Harris const.



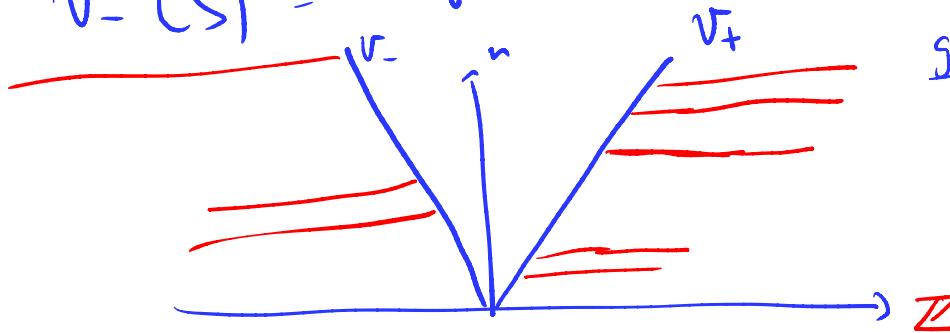
This provides a family of coupled random walks with a monotonicity property if we  $(X_n^y)_{n \geq 0}$  at  $y$ ,

then  $X_n^y \leq X_n^{y'}$  for  $n \geq 0$  iff  $y \leq y'$

### I) Good candidates for the speed

$$v_+(s) = \inf \{ v \in \mathbb{R} : \lim_{n \rightarrow \infty} P^s(X_n \geq v \cdot n) = 0 \}$$

$$v_-(s) = \sup \{ v \in \mathbb{R} : \lim_{n \rightarrow \infty} P^s(X_n \leq v \cdot n) = 0 \}$$



good:  $v_{\pm}$  exist.

bad: \*

\* non-quantitative  
\* we do not even know  
that  $v_- \leq v_+$

## II) Idea to obtain a LLN

1) Prob. to go faster than  $v_+(s)v_0$  (nearf. slower than  $v_-(s)$ )  
decays fast. with  $n$ .

2)  $v_- = v_+(s)$

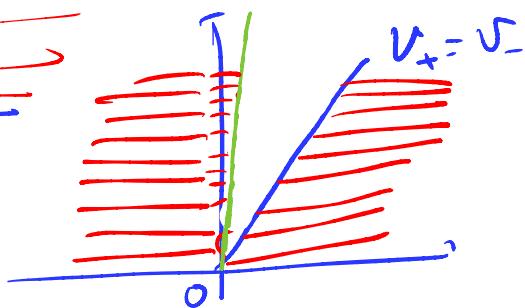
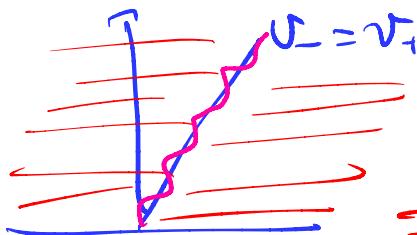
3) Assume  $v_+ = v_- > 0$

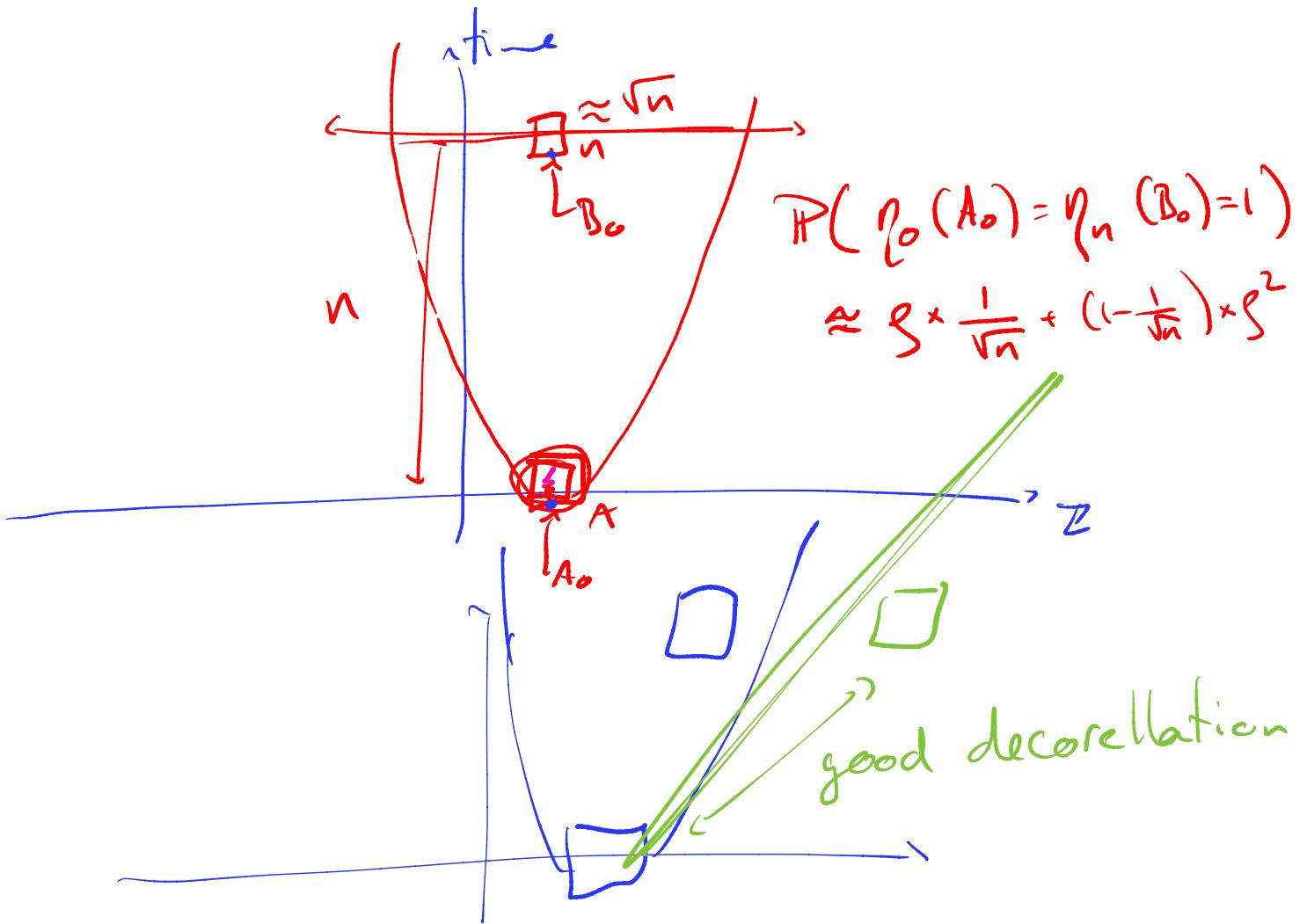
Prove  $P(X_n \leq \varepsilon \cdot n) \rightarrow 0$  fast

Using that, we will be

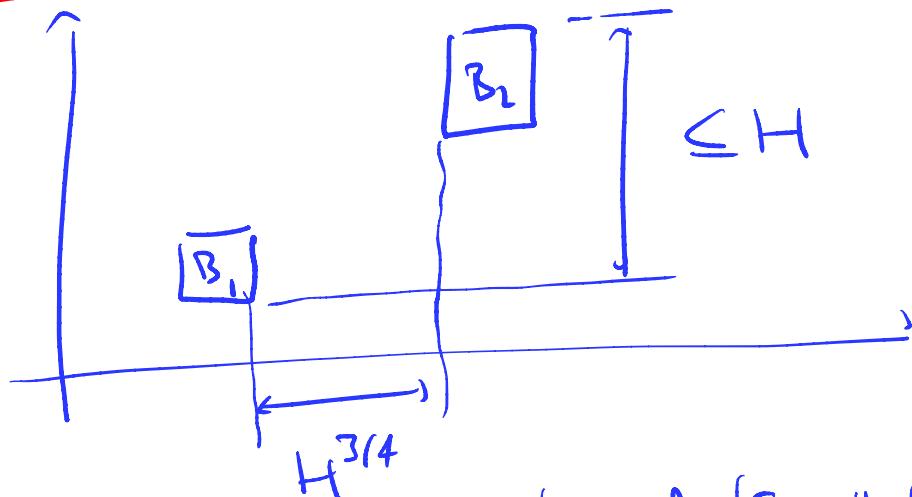
able to use a regeneration structure

$\Rightarrow$  LLN





## Lateral decoupling



$H \delta_1, \delta_2$  non-negative fct<sup>o</sup>,  $\| \cdot \|_{\infty} \leq 1$ , supported  
on  $B_1$  &  $B_2$  respective

$$Core_s(\delta_1, \delta_2) \leq c_1(s) e^{-H^{1/4}}.$$

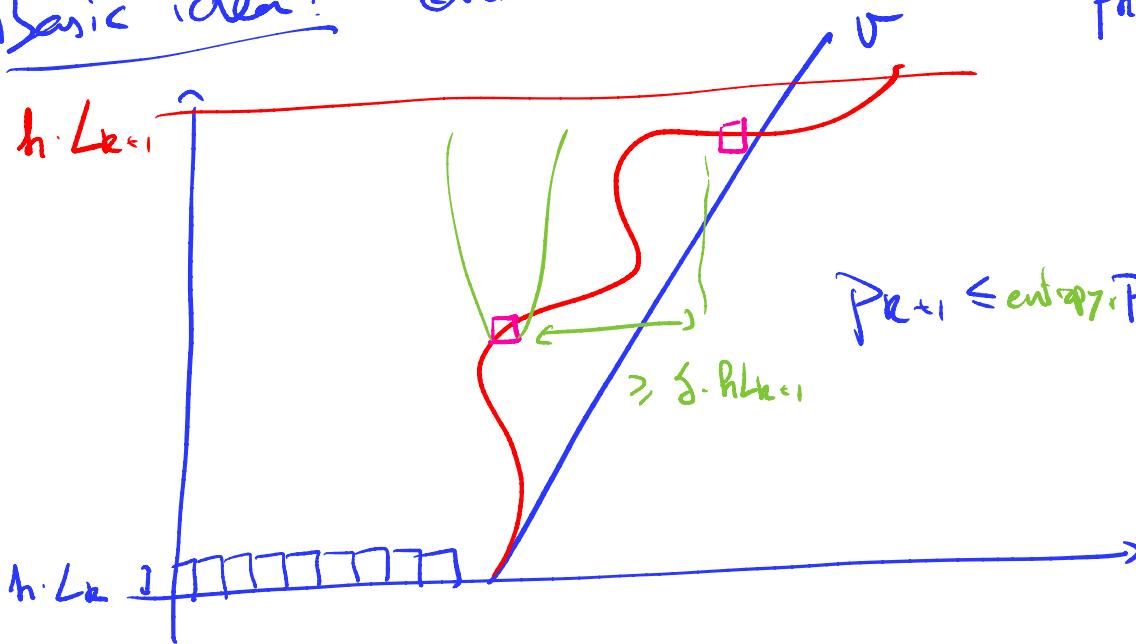
Very useful to control the probab. of events of the form  $\{X_n \geq v \cdot n\}$  with  $v > 0$ .

We will use renormalisation.

Basic idea: Control  $P(X_n \geq v \cdot n)$

$v > 0$

$$P_n = P(X_{hL_n} \geq h \cdot L_n \cdot v)$$



$$P_{k+1} \leq \text{entropy} \cdot P_k^2 + \underbrace{\text{error}}_{\substack{\text{control} \\ \text{via lateral} \\ \text{decoupling}}} \cdot h \cdot L_{k+1}$$

\*The following choice works:

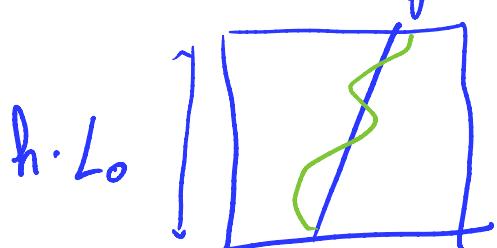
$$L_0 = 10^{10}, \quad L_{k+1} \approx L_k^{5/4} = L_k \cdot L_k^{1/4} \text{ grows super-exponentially.}$$

Assume that  $P_k \leq e^{-4(\log L_k)^{3/2}}$  ← very small in  $L_k$   
 $h \geq 1$ .  
(smaller than polynomial)

Assume

$$\begin{aligned} P_{k+1} &\leq L_k \times P_k^2 + C \cdot e^{-L_{k+1}^{1/4}} \\ &\leq \dots \leq e^{-4(\log L_{k+1})^{3/2}} \\ &\quad \leq \dots \leq e^{-4(\log L_0)^{3/2}} \end{aligned}$$

We need to prove  $P_0 \leq e^{-4(\log L_0)^{3/2}}$



$$\begin{aligned} \text{choose } v &> V_t(s) + \varepsilon \\ \Rightarrow \liminf P(X_n > v \cdot n) &= 0 \end{aligned}$$

Choose  $h$   
 so that:  $P_0 = \mathbb{P}^s(X_{h \cdot L_h} = (h \cdot L_h) \cdot v) \leq \underbrace{e^{-4(\log h)^{3/2}}}_{\text{does not depend on } h.}$

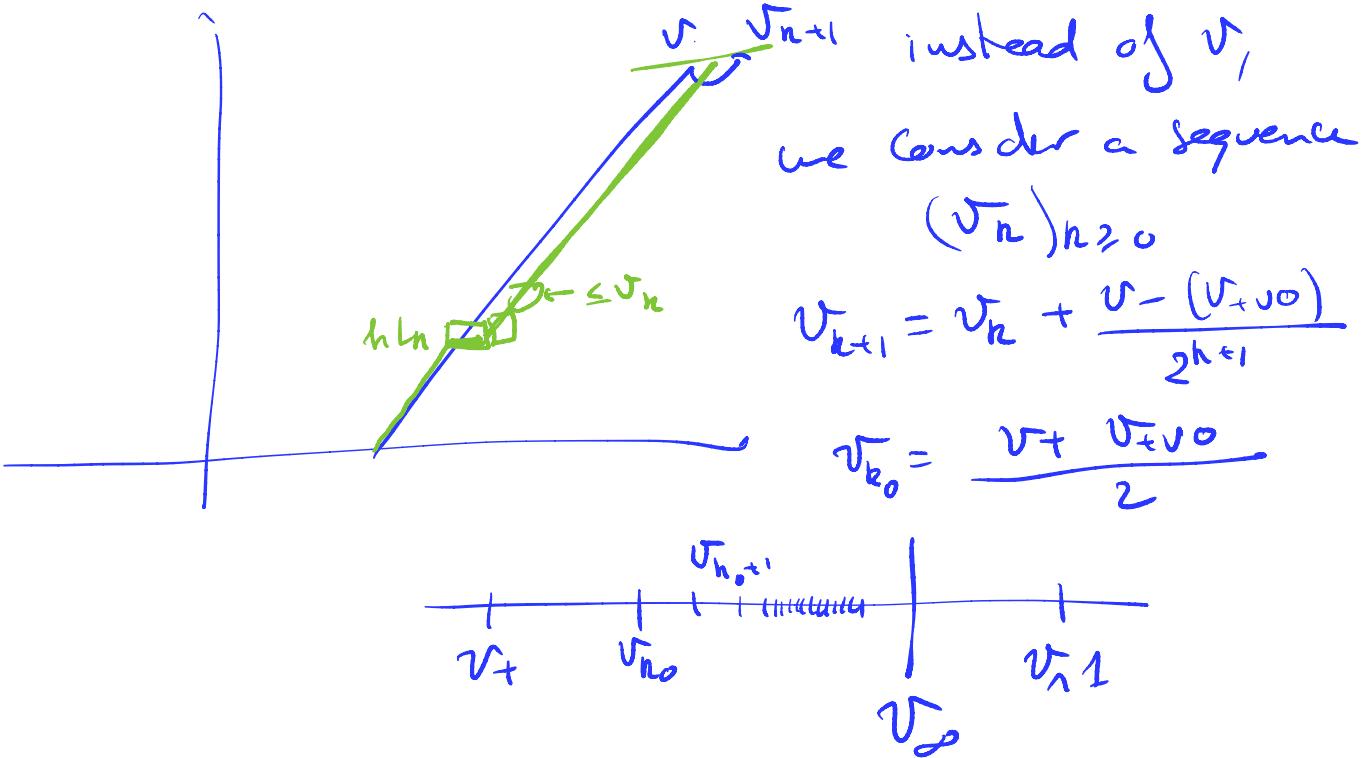
"seed estimating / triggering"  
 + recursive step.

$$\Rightarrow \mathbb{P}(X_{hL_h} \geq hL_h \cdot v) \leq e^{-4(\log h)^{3/2}}, \forall h \geq 0.$$

for  $v > v_+(s)$ .

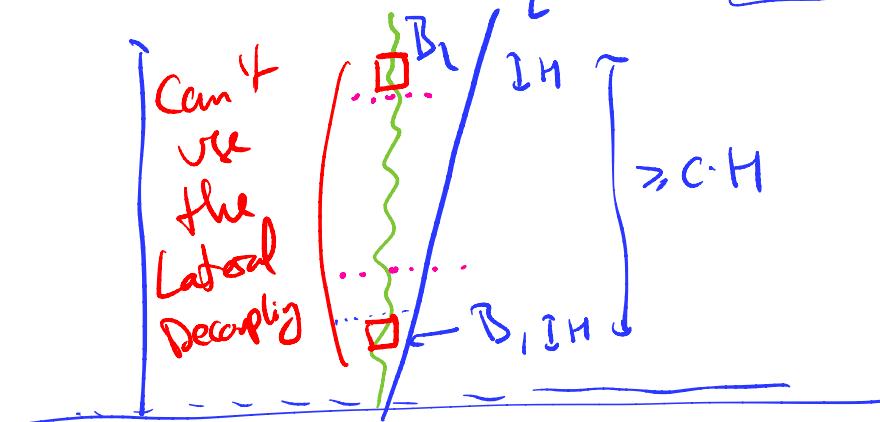
- To obtain the results for all  $n \geq 0$ , it's a straightforward interpolation argument.

$$\Rightarrow \mathbb{P}^s(X_n > (v_+(s) + \varepsilon) \cdot n) \leq e^{-(\log(n))^{3/2}} \quad \underline{\forall n \geq 0}.$$



For the third step,  $\mathbb{P}^s(X_n \leq \varepsilon \cdot n) \rightarrow 0$

when  $V_t(s) = \mathbb{E}(s) > 0$

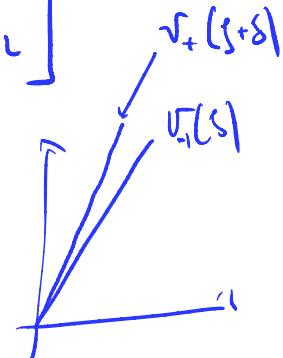


$$\varepsilon_H > \frac{1}{H^{1/b}}$$

For decoupling, we use Sprinkling

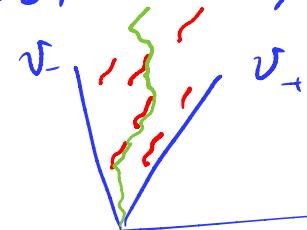
$$\mathbb{E}^{s+\varepsilon_H}[f_1 f_2] \leq \mathbb{E}^{s+\varepsilon_H}[f_1] \times \mathbb{E}^s[f_2]$$

$f_1, f_2 \in [0, 1]$  non-increasing

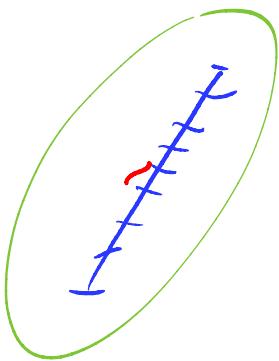


Second step:  $v_-(s) = v_+(\xi)$

Assume  $v_- < v_+$



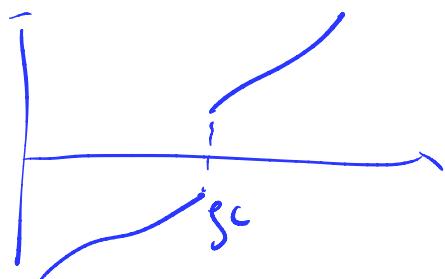
~ good chance to go at speed close to  $v_+$   
~ this will create delays that prevent the walk from approaching, ever,  $v_-$ .  
That's a contradiction



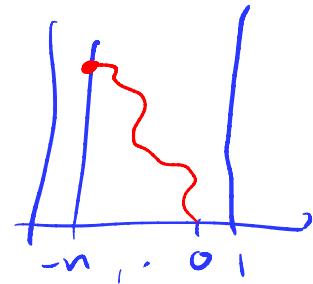
Th (Conchon-Kerjan, K., Rodriguez , '24 )

$$\beta = \beta_+ = \beta_-$$

More precisely  $V(\cdot)$  is strictly monotonic



Sharpness result



Cor:  $B_n = \{X \text{ hits } -n \text{ before } 1\}$

$$1. P^s(B_n) \leq C_1 e^{-(\ln(n))^{3/2}}, \forall n \geq 0, \forall s > s_c,$$

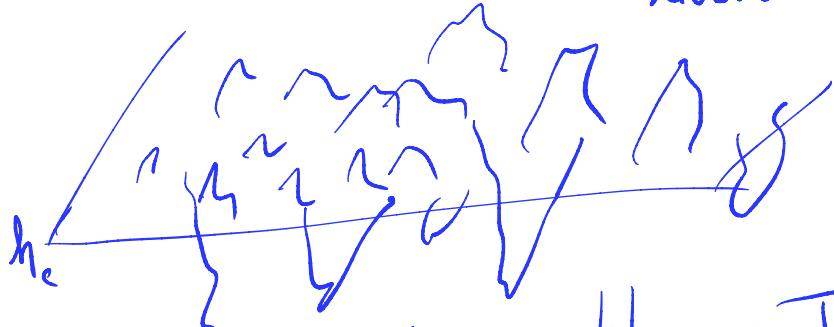
$$2. P^s(B_n) \geq C_2, \forall n \geq 0, \forall s < s_c,$$

critical case  $g_c$  is hard.  $\mathbb{P}^{\text{sc}}(\mathcal{B}_n)??$

\* Our proof strategy is inspired by:

\* Duminil-Copin, Goswami, Rodriguez & Severo ('23)

\* level set percolation for the GFF.  
Gaussian Free Field



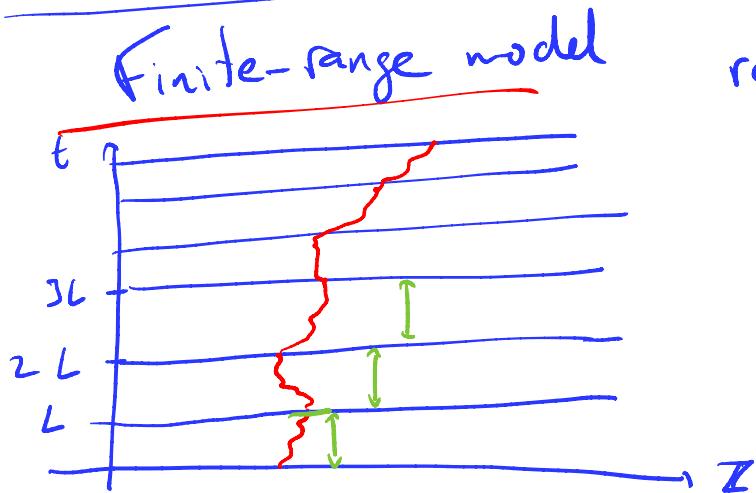
Sane authors + Teixeira ('24)

Random interlacement



## Strategy:

1. Define a finite-range model with good sharpness
2. Compare the finite-range model to the full-range model, via interpolation scheme



refresh the environment every  $L$  steps  
The increment  $(X_{(n+1)L}^{S,L} - X_{nL}^{S,L})_{n \geq 0}$   
are i.i.d.

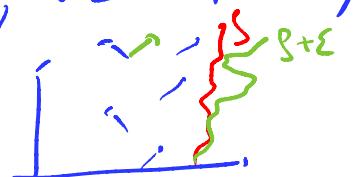
$$\frac{X_n^{S,L}}{n} \rightarrow U_L(S) = \frac{E[X_L^S]}{L}$$

Prop. Approx:  $\forall \gamma \in (0,1)$ ,  $\forall L \text{ large}$ ,  $\alpha_L = \frac{\log^{1+\gamma} L}{L}$

$$\left[ V_L\left(\beta - \frac{1}{\log L}\right) - \alpha_L \leq V(\beta) \leq V_L\left(\beta + \frac{1}{\log L}\right) + \alpha_L \right]$$

Prop. Quantitative Non.:  $\forall \gamma \in (0,1)$ ,  $\forall \epsilon \in (0,1)$ ,  $\forall L \geq L_1(\beta, \epsilon)$

$$\left[ V_L(\beta + \epsilon) - V_L(\beta) \geq 3 \cdot \alpha_L \right]$$

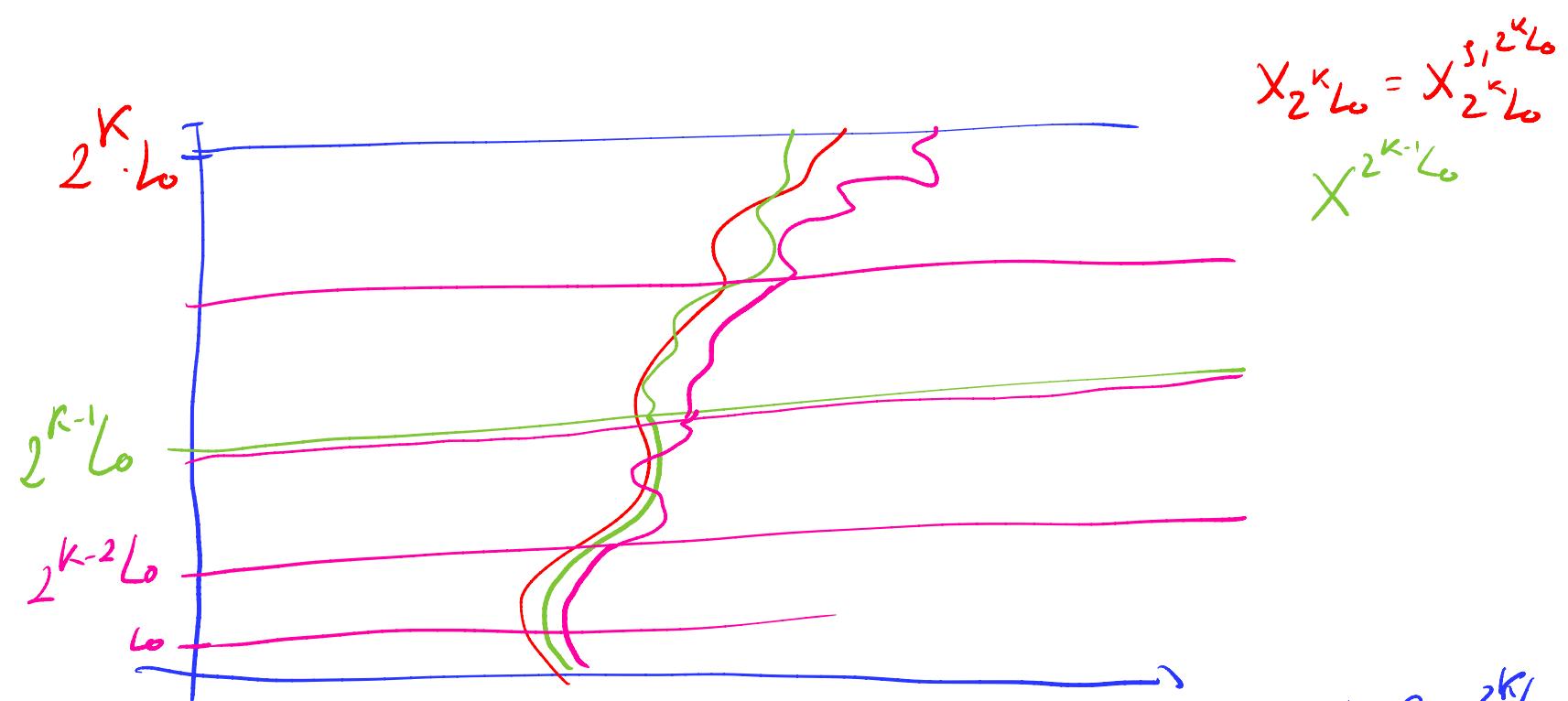


Proof of Th:  
 $V(\beta + 3\epsilon) \stackrel{\text{Approx.}}{\geq} V_L\left(\beta + 3\epsilon - \frac{1}{\log L}\right) - \alpha_L$

$$\geq V_L(\beta + 2\epsilon) - \alpha_L$$

$$\stackrel{\text{Q.N.}}{\geq} V_L(\beta + \epsilon) + 2\alpha_L \stackrel{\text{Approx.}}{\geq} V_L\left(\beta + \frac{1}{\log L}\right) + 2\alpha_L$$

$$\geq V(\beta) + \alpha_L > V(\beta)_{\#}$$



To control the discrepancy between  $X$  &  $X^{2^k L_0}$   
 we need to control the discrepancy between  
 $X^{2 \cdot 2^k L_0}$  &  $X^{2^k L_0}$  for  $h=0, 1, K.$

Compare

$X^{2L, s}$  to  $X^{L, s + \varepsilon_L}$

