

Kinetically constrained models

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European Research Council
Established by the European Commission

YEP 2021

Overview of Lecture 1

- Our first KCM: the FA-2f model
- More examples: the most popular KCM
- Motivations from physics: the liquids/glass transition
- A related deterministic dynamics: Bootstrap Percolation
- Ergodicity and mixing for KCM
- Spectral gap, persistence and mean infection time
- A first tool to upper bound time-scales: the BC technique

Overview of Lecture 2

- East model: scaling for $q \downarrow 0$
- FA-1f model: scaling for $q \downarrow 0$
- The general definition of KCM
- Universality results in $d = 2$:
 - universality classes and results for BP
 - universality classes and results for KCM
 - open issue: the case of sub-critical KCM and BP models

Overview of Lecture 3

- Sharp threshold for FA-2f
 - results
 - heuristics
 - sketch of the proof
- Out of equilibrium
 - key questions
 - results for East model
 - partial results for FA-1f
 - open issues
 - more on East model: aging

Fredrickson Andersen 2 spin facilitated model (FA-2f)

An interacting particle system on $\{0, 1\}^{\mathbb{Z}^d}$, $d \geq 2$.

0=empty, 1=occupied.

Dynamics: **birth and death of particles**

- Fix a parameter $q \in [0, 1]$
- at rate 1 each site gets a proposal to update its state to empty at rate q and to occupied at rate $1 - q$.
- the proposal is accepted **iff** the site has **at least 2 empty nearest neighbours** = *iff the kinetic constraint is satisfied*

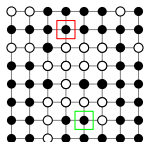
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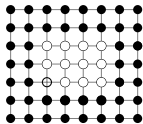
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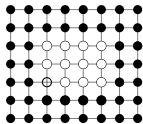


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Several IPS tools fail → new tools needed!

Let's change the constraint: other popular KCM

x can be updated iff ...

- **FA-*j*f:**
... there are at least j empty sites in $\{x \pm \vec{e}_1, \dots, x \pm \vec{e}_d\}$
- **East model:**
... there is at least 1 empty site in $\{x + \vec{e}_1, \dots, x + \vec{e}_d\}$
- **North-East model ($d = 2$):**
... both $x + \vec{e}_1$ and $x + \vec{e}_2$ are empty
- **Duarte model ($d = 2$):**
... there are at least 2 empty sites in $\{x + \vec{e}_2, x - \vec{e}_1, x - \vec{e}_2\}$

KCM: motivations from physics

Introduced in the '80's to model the **liquid/glass transition**

- **major open problem** in condensed matter physics;
- **sharp divergence of timescales**;
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⇒ kinetic constraints of KCM dynamics mimic **cage effect** :
if temperature is lowered free volume shrinks ($q \leftrightarrow e^{-1/T}$)

⇒ **trivial equilibrium** and yet sharp divergence of timescales
when $q \downarrow 0$, aging, heterogeneities, ... → **glassy dynamics**

⇒ **Key question**: how do KCM time-scales diverge for $q \downarrow 0$?

⇒ Sharp divergence → **numerical simulations do not give clear-cut answers, some of the conjectures were wrong!**

Some notation

- $\Omega := \{0, 1\}^{\mathbb{Z}^d}$ is the configuration space
- for $\sigma \in \Omega$ and $x \in \mathbb{Z}^d$, σ_x is the occupation variable at x
- for $\Lambda \subset \mathbb{Z}^d$, $\Omega_\Lambda := \{0, 1\}^\Lambda$ and σ_Λ is the restriction of σ to Λ
- $\mu := \mu^{(q)}$ = Bernoulli $(1 - q)$ product measure on \mathbb{Z}^d
- $\mu_\Lambda := \mu_\Lambda^{(q)}$ = Bernoulli $(1 - q)$ product measure on Λ
- for $f : \Omega \rightarrow \mathbb{R}$, we let $\mu_\Lambda(f) : \Omega_{\mathbb{Z}^d \setminus \Lambda} \rightarrow \mathbb{R}$ be the mean of f w.r.t. μ_Λ with the other variable held fixed
- analogous definition for $\text{Var}_\Lambda(f)$

Formal definition of the Markov process

The **generator** acts on local functions $f : \Omega \rightarrow \mathbb{R}$ as

$$\begin{aligned}\mathcal{L}f(\sigma) &:= \sum_{x \in \mathbb{Z}^d} c_x(\sigma)(\mu_x(f) - f(\sigma)) = \\ &= \sum_{x \in \mathbb{Z}^d} c_x(\sigma)(q\sigma_x + (1-q)(1-\sigma_x))(f(\sigma^x) - f(\sigma))\end{aligned}$$

with

$$\begin{aligned}\sigma^x(y) &:= \begin{cases} \sigma(y) & \text{if } y \neq x \\ 1 - \sigma(x) & \text{if } y = x \end{cases} \\ c_x(\sigma) &:= \begin{cases} 1 & \text{if there constraint is satisfied at } x \\ 0 & \text{otherwise} \end{cases}\end{aligned}$$

The corresponding **Dirichlet form** is:

$$\mathcal{D}(f) := -\mu(f \cdot \mathcal{L}f) = \sum_{x \in \mathbb{Z}^d} \mu(c_x \text{Var}_x(f)).$$

Focus questions

Q. *Is μ ergodic for the infinite volume process? Is it also mixing? And if so, how fast does convergence to equilibrium in $L^2(\mu)$ occur?*

Recall that, if we denote by P_t is the Markov semigroup,

- μ is ergodic if for all $f \in L_2(\mu)$ the condition $P_t f = f \quad \forall t \geq 0$ implies f constant a.s. in μ
- μ is mixing if $\forall f, g \in L^2(\mu)$ it holds $\lim_{t \rightarrow \infty} \mu(f P_t g) = \mu(f)\mu(g)$.

Thus mixing is stronger than ergodicity.

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\Rightarrow To answer the above questions we should first introduce a related model: Bootstrap Percolation (BP)

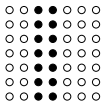
2-neighbour Bootstrap Percolation

- At time $t = 0$ sites are i.i.d., empty with probability q , occupied with probability $1 - q$
- At time $t = 1$
 - each empty site remains empty
 - each occupied site is emptied **iff** it has at least 2 empty n.n.
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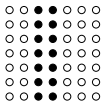
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- BP blocked clusters \leftrightarrow blocked particles under FA-2f
- BP is a discrete time deterministic *monotone* dynamics
→ easier to study

Critical density and Infection time

- Will the whole lattice become empty?
- $q_c := \inf\{q \in [0, 1] : \mu_q(\text{origin is emptied eventually}) = 1\}$
- How many steps are needed to empty the origin?
- $\tau^{\text{BP}}(q) := \mu_q(\text{first time at which origin is empty})$

Critical density and Infection time

- Will the whole lattice become empty?
→ Yes $\forall q > 0$ (Van Enter '87)
- $q_c := \inf\{q \in [0, 1] : \mu_q(\text{origin is emptied eventually}) = 1\}$
→ $q_c = 0$
- How many steps are needed to empty the origin?
- $\tau^{\text{BP}}(q) := \mu_q(\text{first time at which origin is empty})$

$$\text{for } q \downarrow 0 \text{ w.h.p. } \tau_0^{\text{BP}} = \exp\left(\frac{\lambda(d)}{q^{1/(d-1)}}(1 - o(1))\right)$$

- scaling (Aizenmann, Lebowitz '88)
- $\lambda(2) = \pi^2/18$ (Holroyd '08)
- $\lambda(d) = \dots$ $d > 2$ (Balogh, Bollobas, Duminil-Copin, Morris '12)

Bootstrap percolation

- Define analogously the BP processes corresponding to the constraints of FA- j f, Duarte, North-East and East
- $q_c = 0$ for East, Duarte and FA- j f $\forall j \in [1, d]$
- $q_c = 1 - p_c^{OP}$ for North-East
- for $q \downarrow 0$, w.h.p. $\tau_0^{\text{BP}} \sim q^{-1/d}$ for FA-1f and East
- the scalings for FA- j f with $j > 1$ and for Duarte model are more complicate and diverge more rapidly as $q \downarrow 0$ (see Lecture 2)

KCM: ergodicity and mixing

Theorem (Cancrini, Martinelli, Roberto, C.T. '08)

- (i) $\forall q > q_c$, μ is mixing (and therefore ergodic);
- (ii) $\forall q < q_c$, μ is not ergodic (and therefore not mixing).

Sketch of the proof

- for $q > q_c$ we prove that 0 is a simple eigenvalue of \mathcal{L} .
Key ingredient: fix $x \in \mathbb{Z}^d$ and $\sigma \sim \mu$, then μ -a.s. there exists a *legal path* from σ to σ^x ;
- for $q < q_c$ blocked structures percolate
 $\rightarrow f := 1_{\mathcal{E}}$ is left invariant by the dynamics and it is not constant a.s. w.r.t. μ where

$$\mathcal{E} := \{\eta : \text{the origin cannot be emptied by BP}\}.$$

Spectral gap and relaxation time

$$\text{gap} := \inf_{\substack{f \in \text{Dom}(\mathcal{L}) \\ \text{Var}(f) \neq 0}} \frac{\mathcal{D}(f)}{\text{Var}(f)}$$

i.e. $\text{gap} = T_{\text{rel}}^{-1}$, where T_{rel} is the smallest constant such that

$$\text{Var}(f) \leq T_{\text{rel}} \mathcal{D}(f) \quad \forall f$$

Thus, if $\text{gap} > 0$, it holds

$$\text{Var}(P_t f) = \mu(f P_t f) - \mu(f)^2 \leq \exp(-2 t \text{gap}) \text{Var}(f) \quad \forall f \in L^2(\mu)$$

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Theorem (Cancrini, Martinelli, Roberto, C.T. '08)

For FA-jf, East, North-East and Duarte models it holds

$$\text{gap} > 0 \quad \forall q > q_c$$

Persistence function

$$F(t) := \int d\mu(\sigma) \mathbb{P}_\sigma(\sigma_s(0) = \sigma(0) \forall s \leq t)$$

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Recall that $\forall \mathcal{A} \subset \Omega$ it holds

$$\mathbb{P}_\mu(\tau_{\mathcal{A}} > t) \leq \exp(-t\lambda_{\mathcal{A}})$$

with $\tau_{\mathcal{A}}$ the hitting time of \mathcal{A} , and

$$\lambda_{\mathcal{A}} := \inf \left\{ \mathcal{D}(f) : \mu(f^2) = 1, f \equiv 0 \text{ on } \mathcal{A} \right\}.$$

Thus

$$F(t) = \mathbb{P}(\tau_{\{\sigma(0)=1\}} > t) + \mathbb{P}(\tau_{\{\sigma(0)=0\}} > t) \leq e^{-(1-q)t \text{ gap}} + e^{-qt \text{ gap}}$$

Mean infection time

$\mathbb{E}_\mu(\tau_0)$ with $\tau_0 = \text{hitting time of } \{\sigma(0) = 0\}$.

The upper bound on $F(t)$ implies

$$\mathbb{E}_\mu(\tau_0) \leq (1 + o(q)) \frac{T_{\text{rel}}}{q}$$

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An easy lower bound

For any KCM there exists $\delta > 0$ s.t. for q small enough it holds

$$\mathbb{E}_\mu(\tau_0) \geq \delta \tau^{\text{BP}}(q)$$

General but usually very far from the correct scaling.

Key idea: BP features only infecting moves while KCM has both infecting and healing moves \rightarrow BP infects the origin at least as fast as the corresponding KCM.

East model in $d = 1$: $\text{gap} > 0$ via the BC technique

- $\text{gap} :=$ spectral gap for East on \mathbb{Z}
- for $\Lambda \subset \mathbb{Z}$, $\text{gap}_\Lambda :=$ spectral gap for East on Λ with 0 b.c.
- $\Lambda_k := [0, 2^k]$ and $\gamma_k := 1/\text{gap}_{\Lambda_k}$
- $\text{gap} \geq \inf_k \text{gap}_{\Lambda_k}$

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→ if we prove

$$\gamma_k \leq a_k \gamma_{k-1} \quad \text{with} \quad \prod_{k_0}^{\infty} a_k < \infty \quad \text{for a finite } k_0$$

we have proven $\text{gap} > 0$

First idea

- Devide Λ_k into two B_1, B_2 each of the form Λ_{k-1}
- define an auxiliary block dynamics: ...
- $\rightarrow T_{\text{rel}, \Lambda_k} \leq T_{\text{rel}}^{\text{block}} \max T_{\text{rel}, B_1} T_{\text{rel}, B_2}$
- $T_{\text{rel}}^{\text{block}} = 1$ (product measure) $\rightarrow \gamma_k \leq \gamma_{k-1}$
- ... so easy?!
- $\max T_{\text{rel}, B_1} T_{\text{rel}, B_2} = \infty!$

A general two-site constrained Poincaré inequality

Lemma

- (\mathbb{X}_1, ν_1) and $(\mathbb{X}_2, \nu_2) =$ finite probability spaces
- $(\mathbb{X}, \nu) =$ the associated product space
- $\mathcal{H} \subset \mathbb{X}_2$ with $\nu_2(\mathcal{H}) > 0$.

Then, for any $f : \mathbb{X} \rightarrow \mathbb{R}$, it holds

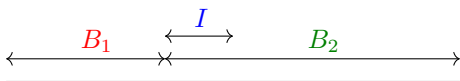
$$\text{Var}_\nu(f) \leq \left(1 - \sqrt{1 - \nu_2(\mathcal{H})}\right)^{-1} \nu \left(1_{\{X_2 \in \mathcal{H}\}} \text{Var}_{\nu_1}(f) + \text{Var}_{\nu_2}(f)\right).$$

- Let $\vec{X} := (X_1, X_2)$. The inequality is an upper bound on T_{rel} for the Markov process reversible w.r.t. ν with generator:

$$\mathcal{L}f(\vec{X}) = 1_{\{X_2 \in \mathcal{H}\}} \left[\nu_1(f) - f(\vec{X}) \right] + \left[\nu_2(f) - f(\vec{X}) \right]$$

- easy proof via direct calculation on eigenvectors

East model in $d = 1$: $\text{gap} > 0$ via the BC technique



$$\Lambda_k = [0, 2^k] = B_1 \sqcup B_2, \quad B_1 := [0, 2^{k-1} - 1], \quad B_2 := [2^{k-1}, 2^k],$$

$$I := [2^{k-1}, 2^{k-1} + 2^{k/3}], \quad \mathcal{H} \subset \Omega_{B_2} := \{\eta : \exists \text{ at least one zero in } I\}$$

Use the two-site constrained Poincaré inequality to get

$$\mathbf{Var}_{\Lambda_k}(\mathbf{f}) \leq \epsilon_k \mu_{\Lambda_k} \left(\mathbf{1}_{\mathcal{H}} \mathbf{Var}_{B_1}(\mathbf{f}) + \mathbf{Var}_{B_2}(\mathbf{f}) \right)$$

$$\text{with } \epsilon_k = \left(1 - \sqrt{(1 - q)^{2^{k/3}}} \right)^{-1}$$

East model in $d = 1$: $\text{gap} > 0$ via the BC technique

Our goal: upper bound the r.h.s with $a_k(\text{gap}_{[0,2^k-1]})^{-1} \mathcal{D}_{\Lambda_k}(f)$

$$\text{r.h.s.} := \epsilon_k \mu_{\Lambda_k} (1_{\mathcal{H}} \text{Var}_{B_1}(f) + \text{Var}_{B_2}(f))$$

Via the Poincaré inequality (i.e. the definition of gap) for East:

$$\mu_{\Lambda_k}(\text{Var}_{B_i}(f)) \leq (\text{gap}(\mathcal{L}_{B_i}))^{-1} \sum_{x \in B_i} \mu_{\Lambda_k}(c_{x,B_i} \text{Var}_x(f))$$

$$c_{x,B_i}(\sigma) := \begin{cases} 1 - \sigma_{x+1} & \text{if } x \neq \text{rightmost site of } B_i \\ 1 & \text{otherwise} \end{cases}$$

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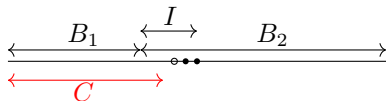
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$$c_{x,B_2}(\sigma) = x, \Lambda_k(\sigma), \text{ but } c_{x,B_1}(\sigma) \geq c_{x,\Lambda_k}(\sigma) !$$

$$\rightarrow \text{r.h.s.} \neq \epsilon_k \gamma_{k-1} \mu(\mathcal{D}_{B_1}(f) + \mathcal{D}_{B_2}(f)) = \epsilon_k \gamma_{k-1} \mathcal{D}_{\Lambda_k}(f)$$

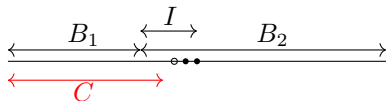
East model in $d = 1$: $gap > 0$ via the BC technique



”Enlarge” Var_{B_1} by convexity to the random interval $C := \dots$

$$c_{x,C}(\sigma)1_{\mathcal{H}} = c_{x,\Lambda_k}(\sigma)1_{\mathcal{H}} \quad \forall x \in C, \sigma \in \Omega_{\Lambda_k}$$

East model in $d = 1$: $\text{gap} > 0$ via the BC technique

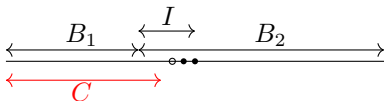


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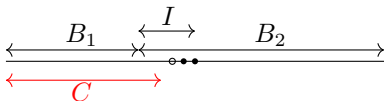
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$$\Rightarrow \text{Var}_{\Lambda_k} \leq \epsilon_k \gamma_{k-1} \mathcal{D}_{\Lambda_k} + \sum_{x \in I} \mu_{\Lambda_k}(c_{x,\Lambda_k} \text{Var}_x(f))$$

- Technical point: ”move” I and average over its positions
- use the variational definition of the spectral gap

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$$\Rightarrow \gamma_{\mathbf{k}} \leq \mathbf{a}_{\mathbf{k}} \gamma_{\mathbf{k}-1} \quad \text{with} \quad \prod_{\mathbf{k}_0}^{\infty} \mathbf{a}_{\mathbf{k}} < \infty \quad \text{for a finite } \mathbf{k}_0$$

East model in $d = 1$: $\text{gap} > 0$ via the BC technique

Theorem (Cancrini, Martinelli, Roberto, C.T '08)

For all $\delta > 0$ there exists C_δ s.t.

$$T_{\text{rel}} = \frac{1}{\text{gap}} \leq C_\delta \exp\left(\frac{|\log q|^2}{(2 - \delta) \log 2}\right)$$

East model in $d = 1$: $\text{gap} > 0$ via the BC technique

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Remark

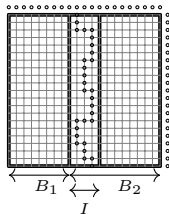
- The conjectures in physics were incorrect, claiming

$$T_{\text{rel}} \sim \exp\left(\frac{|\log q|^2}{\log 2}\right)$$

- The additional factor $1/2$ is due to the fact that (unexpectedly!) energy and entropy contributions are of the same order (more on Lecture 2)

Can BC be used to upper bound T_{rel} for all KCM?

Consider FA-2f on \mathbb{Z}^2



$$\Lambda_k := [1, 2^k] \times [1, 2^k]$$

$$I := [2^{k-1}, 2^{k-1} + 2^{k/3}] \times [1, 2^k]$$

$$\mathcal{H} := \{\eta : \exists \text{ top-bottom empty crossing in } I\}$$

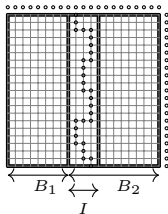
Following the lines of the proof for East on \mathbb{Z} we get

$$\gamma_k := \text{gap}_{\Lambda_k}^{-1} \leq a_k \gamma_{k-1}$$

with $a_k := \left(1 - \sqrt{1 - \mu(\mathcal{H})}\right)^{-1} \rightarrow \prod_{k_0}^{\infty} a_k < 1$ iff $q > q_c^{\text{site perc.}}$

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→ BC cannot be used (alone) to upper bound T_{rel} for all $q > 0!$
Renormalisation + other tools...

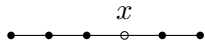
East model $d = 1$: combinatorics

Constraint = to update a site we need its right neighbour empty

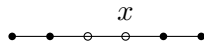
East model $d = 1$: combinatorics

Constraint = to update a site we need its right neighbour empty

- If we start from a single vacancy



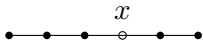
and we can create 1 zero we reach only



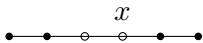
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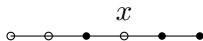
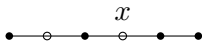
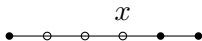
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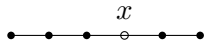
- if we can create up to 2 simultaneous additional zeros we reach also:



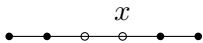
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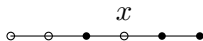
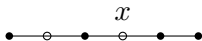
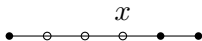
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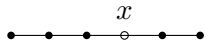


- if we can create up to n simultaneous additional zeros
 - one of the configurations that we can reach has its **leftmost vacancy at $x - (2^n - 1)$** ;
 - all the others have leftmost vacancy in $[x, x - (2^n - 1)]$

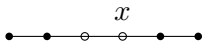
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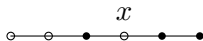
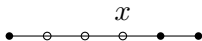
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⇒ the East model has **logarithmic energy barriers**
[Evans Sollich '99, see also Chung Diaconis Graham '01]

East model $d = 1$: scaling for $q \downarrow 0$

- The first vacancy at the left of origin is at $\ell \sim 1/q$
- Trivially, $\tau_0^{\text{BP}}(q) \sim 1/q$
- $\mathbb{E}_{\mu_q}(\tau_0) \sim$ time to create $\log_2(\ell)$ empty sites
- $\rightarrow \mathbb{E}_{\mu_q}(\tau_0) = 1/q^{\Theta(1)|\log q|}$ [Aldous, Diaconis JSP '02]

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- $\rightarrow \mathbb{E}_{\mu_q}(\tau_0) = 1/q^{\Theta(1)|\log q|}$ [Aldous, Diaconis JSP '02]
- Sharp result (taking entropy into account) in $d \geq 1$

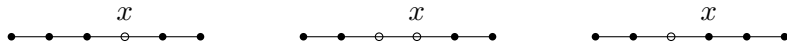
$$\lim_{q \rightarrow 0} \frac{\log \mathbb{E}_{\mu_q}(\tau_0)}{|\log q|^2} = (2d \log 2)^{-1}$$

[Cancrini, Martinelli, Roberto, C.T. PTRF '08] for $d = 1$

[Chleboun, Faggionato, Martinelli AoP '16] for $d \geq 2$

FA-1f: scaling for $q \downarrow 0$

Constraint = to be update we need an empty nearest neighbour



- a vacancy can move of one step by creating one additional vacancy $\rightarrow \sim$ r.w. of rate $q^{-1} \rightarrow$ **non-cooperative!**

$$\rightarrow d = 1 \quad \mathbb{E}_{\mu_q}(\tau_0) \sim q^{-1}(1/q)^2 = q^{-3};$$

$$\rightarrow d = 2 \quad q^{-2} \leq \mathbb{E}_{\mu_q}(\tau_0) \leq q^{-2}|\log q|;$$

$$\rightarrow d \geq 3 \quad \mathbb{E}_{\mu_q}(\tau_0) \sim q^{-1}(1/q^{1/d})^d = q^{-2}$$

[Cancrini, Martinelli, Roberto, C.T. PTRF '08 + Shapira JSP '20]

KCM: the general definition

Configurations : $\eta \in \Omega := \{0, 1\}^{\mathbb{Z}^d}$, 0 = empty, 1 = occupied

Fix a **density parameter** $q \in [0, 1]$ and an **update family** \mathcal{U} with

$$\mathcal{U} = \{U_1, \dots, U_m\}, \quad U_i \subset \mathbb{Z}^d \setminus 0, \quad |U_i| < \infty, \quad m < \infty$$

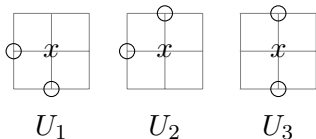
i.e. \mathcal{U} is a finite collection of local neighbourhoods of the origin

Fix $\eta \in \Omega$ and $x \in \mathbb{Z}^d$: **"the constraint is satisfied at x "** iff at least one of the translated sets $U_i + x$ is completely empty

Dynamics: each site with the constraint satisfied is updated to empty at rate q and to occupied at rate $1 - q$

Our examples

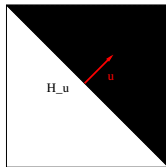
- **FA- j f model:**
 \mathcal{U} = all sets containing j nearest neighbours of the origin
- **East model:** $\mathcal{U} = \{\vec{e}_1, \dots, \vec{e}_d\}$
- **North-East model:** $\mathcal{U} = \{U_1\}$ with $U_1 = \{(0, 1), (1, 0)\}$
- **Duarte model:** $\mathcal{U} = \{U_1, U_2, U_3\}$ with



Universality class of \mathcal{U} in $d = 2$

We need the notion of **stable** and **unstable directions**

- Fix a direction \vec{u}
- Start from a configuration which is
 - completely empty on the half plane perpendicular to \vec{u} in the negative direction (H_u)
 - filled otherwise
- Run the bootstrap dynamics



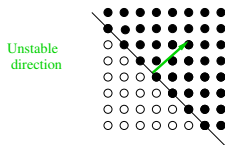
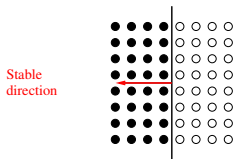
\vec{u} is $\begin{cases} \text{stable} \\ \text{unstable} \end{cases}$ if no other site can be emptied otherwise

Stable and unstable directions: examples

Of course, the stability of a direction depends on \mathcal{U}

Ex. East model:

$\vec{u} = -\vec{e}_1$ is **stable**; $\vec{u} = \vec{e}_1 + \vec{e}_2$ is **unstable**

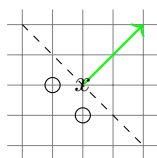


Instead :

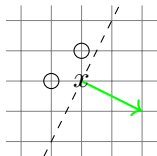
- both directions are unstable for 1-neighbour bootstrap
- both directions are stable for North East

How to easily identify all stable and unstable directions

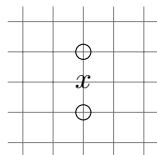
Draw the half planes H_u and $\mathbb{Z}^2 \setminus H_u$ so that the separation line contains the origin. \vec{u} is unstable iff $U_i \subset H_u$ for at least one i



U_1



U_2

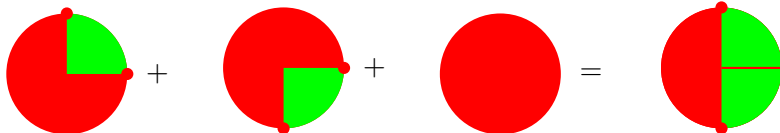


U_3

$$R + R = R$$

$$G + G = G$$

$$G + R = G$$



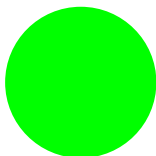
Easy-to-use criterion to determine the class of any \mathcal{U}
 (m simple geometric checks, $m = \#$ of rules)

Supercritical universality class

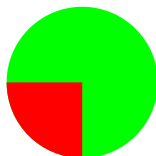
\mathcal{U} is supercritical iff there exists an open semicircle \mathcal{C} which does not contain stable directions.

A supercritical model is

- **rooted** if it has at least 2 non opposite **stable** directions
- **unrooted** otherwise



FA-1f
Unrooted



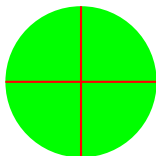
East
Rooted

Critical universality class

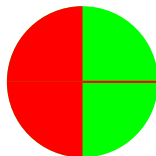
\mathcal{U} is critical if it is not supercritical and there exists an open semicircle \mathcal{C} with only a finite number of stable directions

A critical model is

- finitely critical if it has a finite number of stable directions
- infinitely critical otherwise



FA-2f
Finitely critical



Duarte
Infinitely critical

Subcritical universality class

Two equivalent definitions

\mathcal{U} is subcritical iff it is neither supercritical nor critical

or

\mathcal{U} is subcritical iff each open semicircle has infinite stable directions

$\Rightarrow q_c > 0$: blocked clusters percolate at $q < q_c$

Example: North East model



BP universality results in $d = 2$

Theorem [Bollobás, Smith, Uzzell '15 + Balister, Bollobás, Przykucki, Smith '16 + Bollobás, Duminil-Copin, Morris, Smith '16]

- **Supercritical:** $q_c = 0$, $\tau_0^{\text{BP}}(q) = 1/q^{\Theta(1)}$ w.h.p. as $q \downarrow 0$
- **Critical:** $q_c = 0$, $\tau_0^{\text{BP}}(q) = \exp(|\log q|^{O(1)}/q^\alpha)$ w.h.p. as $q \downarrow 0$
- **Subcritical:** $q_c > 0$

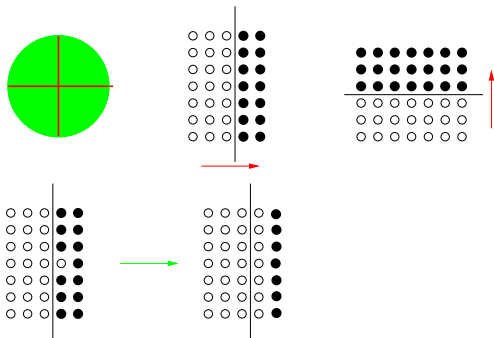
Definition of the "difficulty", α

$\alpha = \min_C \max_{\vec{u} \in C} d(\vec{u})$ with

$d(\vec{u}) =$ minimal number of empty sites to unstabilize \vec{u}

Difficulty: examples

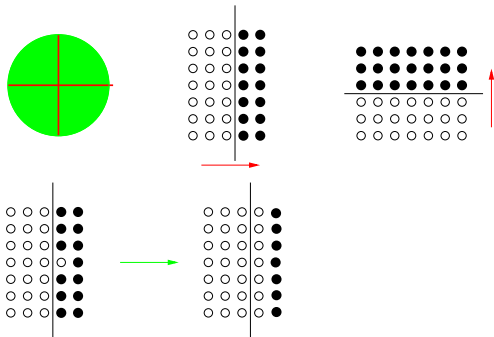
FA-2f :



- \vec{e}_1 is stable and $d(\vec{e}_1) = 1$
- same for $-\vec{e}_1$ and $\pm\vec{e}_2$
- all other directions are unstable

Difficulty: examples

FA-2f :

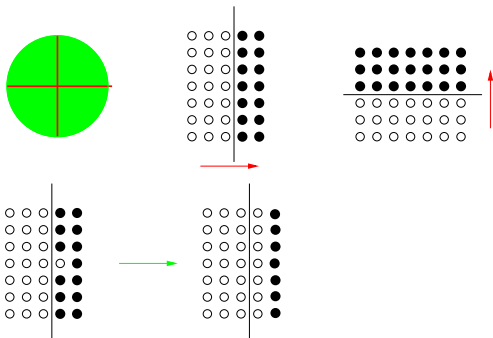


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→ $\alpha = 1$ for FA-2f

Difficulty: examples

FA-2f :



- \vec{e}_1 is stable and $d(\vec{e}_1) = 1$
- same for $-\vec{e}_1$ and $\pm\vec{e}_2$
- all other directions are unstable

→ $\alpha = 1$ for FA-2f

Exercice: check that Duarte model also has difficulty $\alpha = 1$.

KCM: universality results in $d = 2$

Theorem [Martinelli, Morris, C.T. '19; Marêché, Martinelli, C.T. '20, Marêché, Hartarsky, Toninelli '20, Hartarsky, Martinelli, C.T. '21]

① **Supercritical unrooted:** $\tau(q) = \frac{1}{q^{\Theta(1)}}$ and $\tau_0^{\text{BP}} \sim \frac{1}{q^{\Theta(1)}}$ (FA-1f)

② **Supercritical rooted:** $\tau_0(q) = \frac{1}{q^{\Theta(1)|\log q|}} \gg \tau_0^{\text{BP}} = \frac{1}{q^{\Theta(1)}}$ (East)

③ **Finitely critical:** $\tau_0(q)$ and $\tau_0^{\text{BP}} \sim \exp\left(\frac{\Theta(1)(\log q)^{\Theta(1)}}{q^\nu}\right)$ (FA-2f)

④ **Infinitely critical:**

$$\tau(q) = \exp\left(\frac{(\log q)^c}{q^{2\nu}}\right) \gg \tau_0^{\text{BP}} = \exp\left(\frac{(\log q)^c}{q^\nu}\right) \text{ (Duarte)}$$

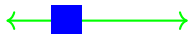
→ For supercritical rooted and infinitely critical models

$$\mathbb{E}_{\mu_q}(\tau_0) \gg \tau_0^{\text{BP}}(q)^\nu \text{ for all } \nu.$$

Hartarsky Marêché '21⁺: log corrections !

Supercritical models

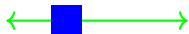
- **Unrooted**: large empty droplet can move back and forth



- renormalise to an FA-1f with effective density $q_{\text{eff}} = q^{\Theta(1)}$
- $\mathbb{E}_{\mu_q}(\tau_0) \sim q^{-\Theta(1)}$

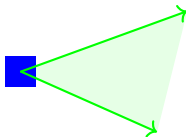
Supercritical models

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- **Rooted**: any empty droplet can move only inside a cone

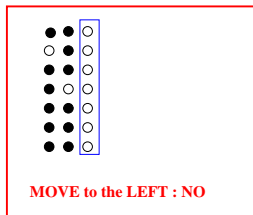
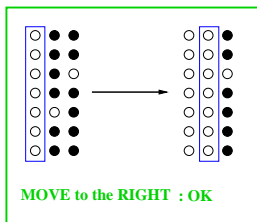


\implies logarithmic energy barriers as for East
[Marêché '20]

- renormalise to an East with effective density $q_{\text{eff}} = q^{\Theta(1)}$
- $\mathbb{E}_{\mu_q}(\tau_0) \sim q_{\text{eff}}^{\Theta(|\log q_{\text{eff}}|)} = e^{\Theta(\log q)^2}$

Duarte model: heuristics

Constraint at x : at least 2 vacancies in $\{x - \vec{e}_1, x + \vec{e}_2, x - \vec{e}_2\}$



An empty segment of length $\ell = 1/q \lfloor \log q \rfloor$ can (typically) create an empty segment to its right, but never to its left!

→ it is a **mobile droplet** with **East-like dynamics** and

$$\text{density } q_{\text{eff}} = q^\ell = e^{-\Theta(\log q)^2/q}$$

Duarte model: heuristics

- nearest empty droplet to the origin is at distance $L \sim q_{\text{eff}}^{-1}$

$$\rightarrow T^{\text{BP}} \sim L = \exp\left(\frac{\Theta(1)|\log q|^2}{q}\right)$$

[Mountford '95]

Duarte model: heuristics

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$$\rightarrow T^{\text{BP}} \sim L = \exp\left(\frac{\Theta(1)|\log q|^2}{q}\right)$$

[Mountford '95]

- Duarte droplets move East like \rightarrow **to empty the origin we have to create $\log(L)$ simultaneous droplets**

$$\rightarrow \mathbb{E}_{\mu_q}(\tau_0) \sim q_{\text{eff}}^{-\log L} \sim \exp\left(\frac{\Theta(1)|\log q|^4}{q^2}\right) \gg T^{\text{BP}}$$

[Martinelli, Morris, C.T. '19 + Marêché, Martinelli, C.T. '20]

The general critical case

- Droplets are empty regions with model dependent shape of size $\ell = q^{-\alpha} |\log q|$ and density $q_{\text{eff}} = q^\ell$

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The general critical case

- Droplets are empty regions with model dependent shape of size $\ell = q^{-\alpha} |\log q|$ and density $q_{\text{eff}} = q^\ell$
- For infinitely critical KCM the droplet motion is East like

$$\rightarrow \tau_0 \sim q_{\text{eff}}^{\Theta(|\log q_{\text{eff}}|)} = \exp\left(\frac{|\log q|^{O(1)}}{q^{2\alpha}}\right)$$

- For finitely critical KCM the droplet motion is a subtle combination of East on mesoscopic scales ($L \sim q^{-\Theta(1)}$) and FA-1f on macroscopic scales ($\sim q_{\text{eff}}^{-1}$)

$$\rightarrow \tau_0 \sim q_{\text{eff}}^{\Theta(\log L)} = \exp\left(\frac{|\log q|^{O(1)}}{q^\alpha}\right)$$

Upper bound: main obstacles and tools

- droplets move only on a **good environment**
- the **environment evolves** and can "lose its goodness"
- no monotonicity \rightarrow we cannot "freeze" the environment
- the **motion of droplets is not random walk like**
- it is very difficult to use canonical path arguments

Upper bound: main obstacles and tools

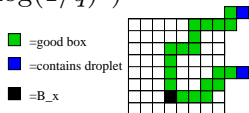
- droplets move only on a **good environment**
 - the **environment evolves** and can "lose its goodness"
 - no monotonicity → we cannot "freeze" the environment
 - the **motion of droplets is not random walk like**
 - it is very difficult to use canonical path arguments
- **a very flexible long range Poincaré inequality**
[Martinelli, C.T. '19]
- **renormalisation**
- **Matryoshka Dolls**: a new technique to compare Dirichlet forms avoiding canonical paths
[Martinelli, Morris, C.T. '19]

Lower bound

- **Key idea:** construct a bottleneck involving $\log(L)$ droplets
- **Main difficulty:** droplets are not "rigid objects"!
- **Solution:** an algorithmic identification of droplets and of an efficient cut-set...

More on the upper bound: the case of FA-2f

- renormalise on $\ell \times \ell$ boxes, $\ell = 1/q \log(1/q)$
- auxiliary long range *block* dynamics:
put equilibrium on box B_x at rate 1 iff it belongs to a good cluster with two droplets at distance at most $L = \exp(1/q \log(1/q)^2)$



- establish a **general long range Poincaré inequality** that yields $T_{rel}^{aux} = O(1)$
- use canonical paths for reversible Markov chains or better repeat the same game inside the path on a smaller scale: now the renormalised sites are the columns of the box ... Matryoshka Dolls!

FA2f: sharp threshold

Theorem [Hartarsky, Martinelli, C.T. '20]

As $q \downarrow 0$, w.h.p. for the stationary FA-2f model on \mathbb{Z}^d it holds

$$\tau_0 = \exp\left(\frac{d \times \lambda(d)}{q^{1/(d-1)}}(1 - o(1))\right), \quad d \geq 2$$

the same result holds for $\mathbb{E}_{\mu_q}(\tau_0)$. Thus, w.h.p. $\tau_0 = (\tau_0^{\text{BP}})^{d+o(1)}$.

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Remark

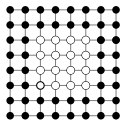
- This is not a corollary of the BP result: the emptying/occupying mechanism of FA-2f has no counterpart in BP!
- We settle contrasting conjectures in physics literature

High level ideas

- Relaxation is driven by the motion of **unlikely large patches of empty sites**, the **mobile droplets**
- droplet density $\rho_D := \exp\left(-\frac{d \times \lambda(d)}{q^{1/d-1}}(1 + o(1))\right)$
droplet length $L_D := \text{poly}(q)$
- **Mobile droplets move in any direction ...**

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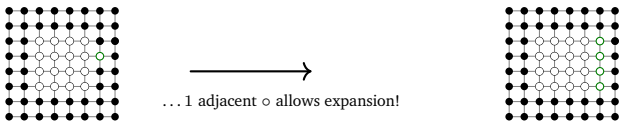
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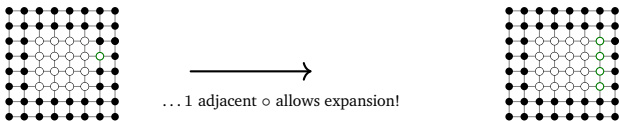
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- Motion requires **few additional empty sites** \rightarrow this **good environment is very likely** for large droplets ($q \downarrow 0$)

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- $\tau_0^{\text{BP}} \sim$ distance of mobile droplet to origin

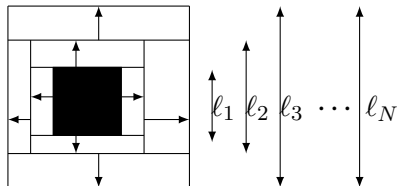
$$\rightarrow \tau_0^{\text{BP}} \sim 1/\rho_D^{1/d} \sim \tau_0^{1/d}$$

How do droplets look like? the $d=2$ case

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- the black square has no double rows fully occupied and one row with no consecutive filled sites \rightarrow it is emptiable
- vertical arrow = no double rows fully occupied
- horizontal arrow = no double columns fully occupied
- $l_n := e^{n\sqrt{q}}/\sqrt{q}$, $N = 8\lceil \log q \rceil/\sqrt{q} \rightarrow l_N = L_D = \text{poly}(q)$

More precisely...

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- Super-good (SG) rectangles:
 - a rectangle of class 0 is SG if it is empty;
 - a rectangle of class n is SG if it contains a SG rectangle R' of class $n - 1$ (the core) AND it satisfies *traversability conditions* elsewhere, i.e. no double column/raw fully occupied.

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Droplets are defined as $\ell_N \times \ell_N$ SG rectangles

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Coalescing + Branching + Simple Exclusion \rightarrow *g*-CBSEP
g for "generalized" (not just 0/1)

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Thanks for your attention!

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- **Advertisement** *Kinetically Constrained Models* C.Toninelli, to appear in *Springer Briefs in Mathematical Physics*

The g -CBSEP chain

- $G = (V, E)$: finite connected graph
- (\mathcal{S}, π) : finite probability space
- $\mathcal{S} = \mathcal{S}_0 \sqcup \mathcal{S}_1$ and $\rho = \pi(\mathcal{S}_1)$
- given $\sigma \in \mathcal{S}^V$, $x \in V$ is *occupied* iff $\sigma_x \in \mathcal{S}_1$
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Theorem [Hartarsky, Martinelli, C.T. '20]

As $\rho \downarrow 0$, $T_{\text{rel}}^{\text{g-CBSEP}} \leq O(\rho^{-1} \log(1/\rho))$

FA- j f model

- for j n-BP for all $d \geq j \geq 2$, w.h.p. it holds

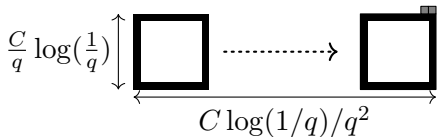
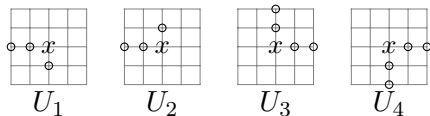
$$\tau_0^{\text{BP}} \sim \exp^{j-1} \left(\frac{\tilde{\lambda}(d, j)}{q^{1/(d-j+1)}} \right)$$

\exp^k = exponential iterated k times (Balogh, Bollobas, Duminil-Copin, Morris '12)

Same scaling for τ_0 (Hartarsky, Martinelli, C.T. in progress)

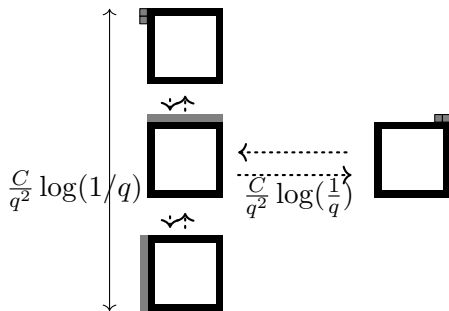
- $j = 1$: $\tau_0^{\text{BP}} = 1/q^{1/d}$, $\tau_0 = 1/q^{\nu(d)}$,
 $\nu(1) = 3$, $\nu(d) = 2$ $d \geq 2$ (log corrections in $d = 2$)
(Cancrini, Roberto, Martinelli, C.T. '08 + Shapira '20)
- $d < j$: $\tau_0 = \tau_0^{\text{BP}} = \infty$ w.h.p. for $q \rightarrow 0$

Finitely critical \mathcal{U} : an example



To move of one step towards \vec{e}_2 the droplet has to move East-like to the right till reaching the first infected pair of empty sites

Finitely critical \mathcal{U} : an example



The move of one step in the $-\vec{e}_1$ direction the droplet has to move in the direction \vec{e}_2 until reaching the first infected pair of empty sites. **A subtle hierarchical combination of East paths...**

East model: why a log barrier?

Start with a single vacancy at the origin

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 $\rightarrow \mathbf{L}(\mathbf{n}) = \mathbf{L}_1(\mathbf{n}) + \mathbf{L}_1(\mathbf{n}-1) + \dots + \mathbf{L}_1(1)$
- 2 to put an isolated 0 at $-L_1(n)$ we should have a 0 at $-L_1(n) + 1$ and remove it using at most $n-1$ vacancies
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 $\rightarrow \mathbf{L}_1(n) = \mathbf{L}(n-1) + 1$

$$\Rightarrow \mathbf{L}(n) = 2^n - 1$$

A general long-range constrained Poincaré inequality

Lemma [Martinelli, C.T. '19]

- (\mathbb{X}, ν) = finite probability space
- $(\tilde{\Omega}, \tilde{\mu}) = (\mathbb{X}^{\mathbb{Z}^d}, \otimes_{x \in \mathbb{Z}^d} \nu_x)$
- for $x \in \mathbb{Z}^d$ let
 - $\mathbb{Z}_{x,\uparrow}^d := x + \{y : y_1 + \dots + y_d > 0\}$
 - $\Delta_x \subset \mathbb{Z}_{x,\uparrow}^d$ be a finite set
 - \mathcal{A}_x be an event depending only on $\{\omega(y)\}_{y \in \Delta_x}$

Assume $\sup_z \sum_{x \in \mathbb{Z}^d, x \cup \Delta_x \ni z} (1 - \mu(\mathcal{A}_x)) < 1/4$, then

$$\text{Var}(f) \leq 4 \sum_{x \in \mathbb{Z}^d} \mu(1_{\mathcal{A}_x} \text{Var}_x(f)) \quad \forall f \text{ local}$$

More general oriented neighborhood and product of characteristic functions possible ...