

# Spread-out limit of the critical points for various statistical-mechanics models

@ Probability and Analysis on Random Structures and Related Topics

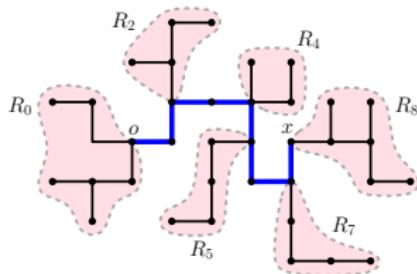
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- Statistical-mechanics models exhibiting phase transitions & critical behavior:
  - self-avoiding walk (SAW),
  - lattice trees (LT) & lattice animals (LA),
  - percolation,
  - oriented percolation (OP) and the contact process (CP).
- The 2-point function  $\tau_p(x)$  & the critical point  $p_c := \sup \{p \geq 0 : \sum_x \tau_p(x) < \infty\}$ .

E.g.,  $\tau_p^{\text{LT}}(x) = \sum_{T \in \mathcal{T}_{o,x}} \left(\frac{p}{|\Lambda|}\right)^{|E_T|},$



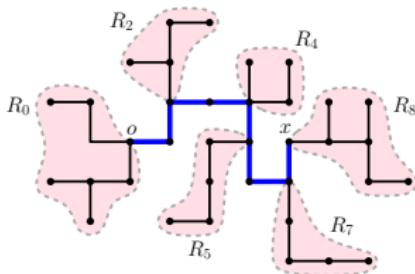
where  $\Lambda = \{x \in \mathbb{Z}^d : 0 < \|x\| \leq L\}$ .

- $\lim_{L \uparrow \infty} p_c = 1$  for SAW, percolation, OP & CP;  $\lim_{L \uparrow \infty} p_c = 1/e$  for LT & LA.
- Want to know: the spread of convergence to those mean-field values.

E.g.,  $\lim_{L \uparrow \infty} \frac{p_c - 1/e}{L^{-d}} \stackrel{d \geq 8}{=} \sum_{n=2}^{\infty} \frac{n+1}{2e} U^{*n}(o) - \begin{cases} 0 & [\text{LT}], \\ \frac{1}{2e} \sum_{n=3}^{\infty} U^{*n}(o) & [\text{LA}]. \end{cases}$

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E.g.,  $\lim_{L \uparrow \infty} \frac{p_c - 1/e}{L^{-d}} \stackrel{d>8}{=} \sum_{n=2}^{\infty} \frac{n+1}{2e} U^{*n}(o) - \begin{cases} 0 & [\text{LT}], \\ \frac{1}{2e^2} \sum_{n=3}^{\infty} U^{*n}(o) & [\text{LA}]. \end{cases}$

- The spread-out edges  $\mathbb{B}_L^d := \{(x, y) \subset \mathbb{Z}^d : x - y \in \Lambda\}$ .
- The 1-step distribution  $D(x) := \frac{1}{|\Lambda|} \mathbb{1}_{\{x \in \Lambda\}}$ .  $\sigma^2 := \sum_{x \in \mathbb{Z}^d} \|x\|^2 D(x) = O(L^2)$ .
- For each  $E \subset \mathbb{B}_L^d$ , the Boltzmann weight  $W_p(E) := \prod_{\{x,y\} \in E} p D(x - y) = \left(\frac{p}{|\Lambda|}\right)^{|E|}$ .
- The 2-point function  $\tau_p(x) := \sum_{E \ni x} W_p(E)$ , where
  - $\blacksquare = \Omega_{o,x}$ : the set of spread-out random-walk (RW) paths between  $o$  and  $x$ ;
  - $\blacksquare = \mathcal{S}_{o,x}$ : the set of spread-out SAWs between  $o$  and  $x$ ;
  - $\blacksquare = \mathcal{T}_{o,x}$ : the set of spread-out LTs containing  $o$  and  $x$ ;
  - $\blacksquare = \mathcal{A}_{o,x}$ : the set of spread-out LAs containing  $o$  and  $x$ .

E.g., since  $\sum_x |\{\omega = (\omega_0, \omega_1, \dots, \omega_n) \in \Omega_{o,x}\}| = |\Lambda|^n$ ,

$$p < 1 \Rightarrow \begin{cases} \chi_p^{\text{RW}} := \sum_x \tau_p^{\text{RW}}(x) = \sum_{n=0}^{\infty} p^n = (1-p)^{-1}, \\ \xi_p^{\text{RW}} := \sqrt{\sum_x \|x\|^2 \frac{\tau_p^{\text{RW}}(x)}{\chi_p^{\text{RW}}}} = \sqrt{\sigma^2 p} (1-p)^{-1/2}, \end{cases}$$

$$\tau_p^{\text{RW}}(x) = \delta_{o,x} + (pD * \tau_p^{\text{RW}})(x) := \delta_{o,x} + \sum_{y \in \mathbb{Z}^d} pD(y) \tau_p^{\text{RW}}(x - y).$$

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- Penrose (1994):  $\exists C L^{-2d/7} \log L \geq p_c^{\text{SAW}} - 1 \geq \begin{cases} 0 & [d \geq 3], \\ \exists c L^{-2} \log L & [d = 2], \\ \exists c' L^{-4/5} & [d = 1]. \end{cases}$
- Madras-Slade (1993):  $d > 4 = d_c^{\text{SAW}} \Rightarrow p_c^{\text{SAW}} = 1 + O(L^{-d}).$
- Hara-Slade (1990) & v.d.Hofstad-Sakai (2005):  
 $d > 6 = d_c^{\text{perc}} \Rightarrow p_c^{\text{perc}} = 1 + O(L^{-d}).$

- v.d.Hofstad-Slade (2002, 2003) & v.d.Hofstad-Sakai (2004):

$$d > 4 = d_c^{\text{OP}} = d_c^{\text{CP}} \Rightarrow p_c^{\text{OP}}, p_c^{\text{CP}} = 1 + O(L^{-d}).$$

- Bramson-Durrett-Swindle (1989):  $p_c^{\text{CP}} - 1 \asymp \lambda_L := \begin{cases} L^{-d} & [d \geq 3], \\ L^{-2} \log L^2 & [d = 2], \\ L^{-2/3} & [d = 1]. \end{cases}$
- Durrett-Perkins (1999):  $\lim_{L \uparrow \infty} \frac{p_c^{\text{CP}} - 1}{\lambda_L} = \begin{cases} \sum_{n=2}^{\infty} U^{*n}(o) & [d \geq 3], \\ \frac{3}{2\pi} & [d = 2], \end{cases}$   
 where  $U$  is the uniform distribution on the  $d$ -dim. ball  $\{x \in \mathbb{R}^d : \|x\| \leq 1\}$ .
- Penrose (1994):  $p_c^{\text{LT}}, p_c^{\text{LA}} = \frac{1}{e} + O(L^{-2d/7} \log L) \quad \forall d \geq 1.$

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- Clisby-Liang-Slade (2007):

$$p_c^{\text{SAW}} = 1 + \frac{1}{2d} + \frac{2}{(2d)^2} + \frac{6}{(2d)^3} + \frac{27}{(2d)^4} + \frac{157}{(2d)^5} + \dots (\text{up to } O(d^{-10})).$$

- Hara-Slade (1995):  $p_c^{\text{perc}} = 1 + \frac{1}{2d} + \frac{7/2}{(2d)^2} + O(d^{-3})$ .
- Cox-Durrett (1983):  $p_c^{\text{OP}} = 1 + O(d^{-2})$ .
- Liggett (1999):  $p_c^{\text{CP}} = 1 + O(d^{-1})$ .
- Miranda-Slade (2013):

$$p_c = \frac{1}{e} + \frac{3}{2e} \frac{1}{2d} + \begin{cases} \frac{115}{24e} \frac{1}{(2d)^2} + o(d^{-2}) & [\text{LT}], \\ \left(\frac{115}{24e} - \frac{1}{2e^2}\right) \frac{1}{(2d)^2} + o(d^{-2}) & [\text{LA}]. \end{cases}$$

### Theorem 1 (v.d.Hofstad-Sakai (2005))

$d > d_c \Rightarrow p_c^{\text{SAW}}, p_c^{\text{perc}}, p_c^{\text{OP}}, p_c^{\text{CP}} = 1 + CL^{-d} + O(L^{-d-1}), \text{ where}$

$$C = \begin{cases} \sum_{n=2}^{\infty} U^{*n}(o) & [\text{SAW \& CP}], \\ \frac{1}{2} \sum_{n=2}^{\infty} U^{*2n}(o) & [\text{OP}], \\ U^{*2}(o) + \sum_{n=3}^{\infty} \frac{n+1}{2} U^{*n}(o) & [\text{percolation}]. \end{cases}$$

### Theorem 2 (Kawamoto-Sakai (2022))

$d > 8 \Rightarrow p_c^{\text{LT}}, p_c^{\text{LA}} = \frac{1}{e} + CL^{-d} + O(L^{-d-1}), \text{ where}$

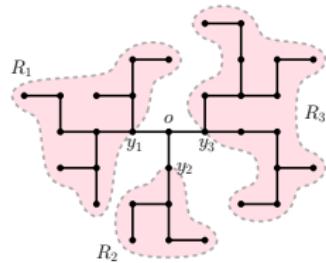
$$C_{\text{LT}} = \sum_{n=2}^{\infty} \frac{n+1}{2e} U^{*n}(o), \quad C_{\text{LA}} = C_{\text{LT}} - \frac{1}{2e^2} \sum_{n=3}^{\infty} U^{*n}(o).$$

① Lace expansion (e.g., Brydges-Spencer (1985), Hara-Slade (1990)):

$$\tau_p = \begin{cases} \delta_o + (pD + \pi_p) * \tau_p & [\text{SAW}], \\ g_p \delta_o + \pi_p + (g_p \delta_o + \pi_p) * pD * \tau_p & [\text{LT}], \\ g_p \delta_o + h_p + \pi_p + (g_p \delta_o + h_p + \pi_p) * pD * \tau_p & [\text{LA}], \end{cases}$$

where

- $\pi_p = \sum_{n=1}^{\infty} (-1)^n \pi_p^{(n)}$ : the model-dependent lace-expansion coefficients,



- $g_p := \sum_{T \in \mathcal{T}_o} W_p(T)$  or  $\sum_{A \in \mathcal{A}_o} W_p(A)$ : the 1-point function,

$$\bullet \quad h_p(x) := (1 - \delta_{o,x}) \sum_{A \in \mathcal{A}_{o,x}} W_p(A) \mathbb{1}_{\{o \iff x \text{ in } A\}}.$$

$$\therefore \chi_p = \hat{\tau}_p := \sum_x \tau_p(x) = \begin{cases} (1 - p - \hat{\pi}_p)^{-1} & [\text{SAW}], \\ ((g_p + \hat{\pi}_p)^{-1} - p)^{-1} & [\text{LT}], \\ ((g_p + \hat{h}_p + \hat{\pi}_p)^{-1} - p)^{-1} & [\text{LA}]. \end{cases}$$

② Hara-v.d.Hofstad-Slade (2003) (& Chen-Sakai (2015)):

$$\text{SAW on } \mathbb{Z}_{L \gg 1}^{d>4} \Rightarrow \begin{cases} \pi_p^{(1)}(x) = \delta_{o,x} (pD * \tau_p)(o), \\ \pi_p^{(n)}(x) = \frac{O(L^{-6}) O(L^{-d})^{n-2}}{(\|x\| \vee L)^{3d-6}} & [\forall n \geq 2, \forall p \leq p_c], \\ \tau_{p_c}(x) = \frac{O(L^{-2})}{(\|x\| \vee L)^{d-2}} & [\forall x \neq o], \\ \chi_p \asymp (p_c - p)^{-1} & [p \uparrow p_c]. \end{cases}$$

$$\begin{aligned} \therefore p_c &= 1 - \hat{\pi}_{p_c} = 1 + \hat{\pi}_{p_c}^{(1)} + O(L^{-2d}) \\ &= 1 + \hat{\pi}_1^{(1)} + \underbrace{(p_c - 1)}_{O(L^{-d})} \underbrace{\partial_p \hat{\pi}_{p_*}^{(1)}}_{O(L^{-d})} + O(L^{-2d}) \\ &= 1 + (D * \tau_1)(o) + O(L^{-2d}) \\ &= 1 + \sum_{n=2}^{\infty} D^{*n}(o) - \underbrace{\sum_x D(x) \sum_{\omega \in \Omega_{o,x} \setminus \mathcal{S}_{o,x}} W_1(\omega)}_{O(L^{-2d})} + O(L^{-2d}). \end{aligned}$$

N.b.,  $\sum_{n=2}^{\infty} D^{*n}(o) = L^{-d} \sum_{n=2}^{\infty} U^{*n}(o) + O(L^{-d-1})$ .

### ③ Hara-v.d.Hofstad-Slade (2003) & Liang (2022):

$$\text{LT \& LA on } \mathbb{Z}_{L \gg 1}^{d>8} \Rightarrow \begin{cases} 1 \leq g_p \leq 4 & [\forall p \leq p_c], \\ \pi_p^{(n)}(x) = \frac{O(L^{-6}) O(L^{-d})^{n-1}}{(\|x\| \vee L)^{2d-6}} & [\forall n \geq 1, \forall p \leq p_c], \\ h_p(x) \leq (1 - \delta_{o,x}) \tau_p(x)^2, \\ \tau_{p_c}(x) = \frac{O(L^{-2})}{(\|x\| \vee L)^{d-2}} & [\forall x \neq o], \\ \chi_p \asymp (p_c - p)^{-1/2} & [p \uparrow p_c]. \end{cases}$$

$$\therefore p_c = \left( g_{p_c} + \hat{h}_{p_c} + \hat{\pi}_{p_c} \right)^{-1} = \frac{1}{g_{p_c}} \left( 1 - \frac{\hat{h}_{p_c} + \hat{\pi}_{p_c}}{g_{p_c}} \right) + O(L^{-2d}).$$

④ Inductive approach: let  $p_1 g_{p_1} = 1$ ,  $\ell_1 = 1$ ,  $\ell_{j+1} = 1 + \frac{\ell_j}{2}$  (so that  $\ell_j \uparrow 2$ ).

$$\bullet 0 \leq 1 - \frac{p_1}{p_c} = - \underbrace{\frac{g_{p_c} - g_{p_1}}{g_{p_1}}}_{\geq 0} - \frac{\hat{h}_{p_c} + \hat{\pi}_{p_c}}{g_{p_1}}. \quad \therefore p_c - p_1 = O(L^{-d\ell_1}).$$

$$\bullet g_{p_c} - g_{p_1} = \left( 1 - \frac{p_1}{p_c} \right) \underbrace{\sum_{T \in \mathcal{T}_o} \sum_{n=1}^{|E_T|} \left( \frac{p_c}{p_1} \right)^n W_{p_1}(T)}_{=: F} = \dots = - \frac{\hat{h}_{p_c} + \hat{\pi}_{p_c}}{g_{p_1} + F} F.$$

$$\therefore p_c - p_1 = p_1 \frac{\hat{\pi}_{p_c}^{(1)} - \hat{h}_{p_c}}{g_{p_1} + F} + O(L^{-2d}). \quad \therefore p_c - p_1 = O(L^{-d\ell_1}) \stackrel{-F \geq \frac{1}{2}\chi_{p_1}}{\Rightarrow} p_c - p_1 = O(L^{-d\ell_{j+1}}).$$

### ③ Hara-v.d.Hofstad-Slade (2003) & Liang (2022):

$$\text{LT \& LA on } \mathbb{Z}_{L \gg 1}^{d>8} \Rightarrow \begin{cases} 1 \leq g_p \leq 4 & [\forall p \leq p_c], \\ \pi_p^{(n)}(x) = \frac{O(L^{-6}) O(L^{-d})^{n-1}}{(\|x\| \vee L)^{2d-6}} & [\forall n \geq 1, \forall p \leq p_c], \\ h_p(x) \leq (1 - \delta_{o,x}) \tau_p(x)^2, \\ \tau_{p_c}(x) = \frac{O(L^{-2})}{(\|x\| \vee L)^{d-2}} & [\forall x \neq o], \\ \chi_p \asymp (p_c - p)^{-1/2} & [p \uparrow p_c]. \end{cases}$$

$$\therefore p_c = (g_{p_c} + \hat{h}_{p_c} + \hat{\pi}_{p_c})^{-1} = \frac{1}{g_{p_c}} \left( 1 - \frac{\hat{h}_{p_c} + \hat{\pi}_{p_c}}{g_{p_c}} \right) + O(L^{-2d}).$$

④ Inductive approach: let  $p_1 g_{p_1} = 1$ ,  $\ell_1 = 1$ ,  $\ell_{j+1} = 1 + \frac{\ell_j}{2}$  (so that  $\ell_j \uparrow 2$ ).

- $0 \leq 1 - \frac{p_1}{p_c} = - \underbrace{\frac{g_{p_c} - g_{p_1}}{g_{p_1}}}_{\geq 0} - \frac{\hat{h}_{p_c} + \hat{\pi}_{p_c}}{g_{p_1}}.$   $\therefore p_c - p_1 = O(L^{-d\ell_1}).$

- $g_{p_c} - g_{p_1} = \left(1 - \frac{p_1}{p_c}\right) \underbrace{\sum_{T \in \mathcal{T}_o} \sum_{n=1}^{|E_T|} \left(\frac{p_c}{p_1}\right)^n W_{p_1}(T)}_{=: F} = \dots = - \frac{\hat{h}_{p_c} + \hat{\pi}_{p_c}}{g_{p_1} + F} F.$

$$\therefore p_c - p_1 = p_1 \frac{\hat{\pi}_{p_c}^{(1)} - \hat{h}_{p_c}}{g_{p_1} + F} + O(L^{-2d}). \quad \therefore p_c - p_1 = O(L^{-d\ell_j}) \stackrel{F \geq \frac{1}{2} \chi_{p_1}}{\Rightarrow} p_c - p_1 = O(L^{-d\ell_{j+1}}).$$

## ⑤ Identifying the dominant contribution to $g_{p_1}$ for LT:

$$g_{p_1} = \text{Diagram with three shaded regions and a central point } o \text{ connected by dashed arcs, with red arrows indicating clockwise flow} = \text{Diagram with two shaded regions and a central point } o \text{ connected by dashed arcs} - \text{Diagram with three shaded regions and a central point } o \text{ connected by dashed arcs, with one region partially shaded.}$$

where

$$\begin{aligned} & \text{Diagram with three shaded regions and a central point } o \text{ connected by dashed arcs} \\ &= \sum_{Y \subset \Lambda} \left( \frac{p_1}{|\Lambda|} \right)^{|Y|} \prod_{y \in Y} \sum_{R_y \in \mathcal{T}_y \setminus \mathcal{T}_o} \left( \frac{p_1}{|\Lambda|} \right)^{|E_{R_y}|} \\ &= \sum_{Y \subset \Lambda} \left( \frac{p_1}{|\Lambda|} \right)^{|Y|} \underbrace{\prod_{y \in Y} \left( \sum_{R_y \in \mathcal{T}_y} \left( \frac{p_1}{|\Lambda|} \right)^{|E_{R_y}|} \right)}_{g_{p_1}} - \underbrace{\sum_{R_y \in \mathcal{T}_{o,y}} \left( \frac{p_1}{|\Lambda|} \right)^{|E_{R_y}|}}_{\tau_{p_1}(y)} \\ &= \sum_{Y \subset \Lambda} \left( \frac{1}{|\Lambda|} \right)^{|Y|} \prod_{y \in Y} \left( 1 - \frac{\tau_{p_1}(y)}{g_{p_1}} \right). \end{aligned}$$

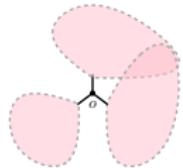
N.b.,  $\sum_{Y \subset \Lambda} \left( \frac{1}{|\Lambda|} \right)^{|Y|} = \left( 1 + \frac{1}{|\Lambda|} \right)^{|\Lambda|} = e \left( 1 - \frac{1}{2|\Lambda|} \right) + O(|\Lambda|^{-2}).$

## ⑥ Identifying sub-dominant contributions to $g_{p_1}$ for LT:

$$\begin{aligned}
 \sum_{Y \neq \emptyset} \left( \frac{1}{|\Lambda|} \right)^{|Y|} \left( \prod_{y \in Y} \left( 1 - \frac{\tau_{p_1}(y)}{g_{p_1}} \right) - 1 \right) &= \sum_{Y \neq \emptyset} \left( \frac{1}{|\Lambda|} \right)^{|Y|} \sum_{\emptyset \neq Z \subset Y} \prod_{y \in Z} \frac{-\tau_{p_1}(y)}{g_{p_1}} \\
 &= \sum_{Z \neq \emptyset} \prod_{y \in Z} \frac{-\tau_{p_1}(y)}{g_{p_1}} \left( \frac{1}{|\Lambda|} \right)^{|Z|} \sum_{Y \supset Z} \left( \frac{1}{|\Lambda|} \right)^{|Y \setminus Z|} \\
 &= \sum_{Z \neq \emptyset} \prod_{y \in Z} \left( \frac{-\tau_{p_1}(y)}{g_{p_1}} D(y) \right) \left( 1 + \frac{1}{|\Lambda|} \right)^{|\Lambda \setminus Z|} \\
 &= \sum_{Z:|Z|=1} \dots \dots \dots + \underbrace{\sum_{Z:|Z| \geq 2} \dots \dots \dots}_{O(L^{-2d})}
 \end{aligned}$$

N.b.,  $\sum_{Z:|Z|=1} \dots = -\left( \frac{\tau_{p_1}}{g_{p_1}} * D \right)(o) \left( 1 + \frac{1}{|\Lambda|} \right)^{|\Lambda|-1} = -e \sum_{n=2}^{\infty} D^{*n}(o) + O(L^{-2d}).$

Similarly,



$$= e \sum_{n=3}^{\infty} \frac{n-1}{2} D^{*n}(o) + O(L^{-2d}).$$

- For SAW, percolation, OP, CP for  $d > d_c$ ,

$$p_c = 1 + CL^{-d} + O(L^{-d-1}), \quad C = \begin{cases} \sum_{n=2}^{\infty} U^{*n}(o) & [\text{SAW \& CP}], \\ \frac{1}{2} \sum_{n=2}^{\infty} U^{*2n}(o) & [\text{OP}], \\ U^{*2}(o) + \sum_{n=3}^{\infty} \frac{n+1}{2} U^{*n}(o) & [\text{percolation}]. \end{cases}$$

- For LT & LA for  $d > 8$ ,

$$p_c = \frac{1}{e} + CL^{-d} + O(L^{-d-1}), \quad C = \sum_{n=2}^{\infty} \frac{n+1}{2e} U^{*n}(o) + \begin{cases} 0 & [\text{LT}], \\ -\frac{1}{2e^2} \sum_{n=3}^{\infty} U^{*n}(o) & [\text{LA}]. \end{cases}$$

- The critical point is a divergent point of  $\chi_p = \sum_x \tau_p(x)$ . E.g., for SAW,  
**Lace expansion**  $\Rightarrow \tau_p = \delta_o + (pD + \pi_p) * \tau_p \Rightarrow \chi_p = (1 - p - \hat{\pi}_p)^{-1}$ .
- $p = e^{-h-X_b}$  ( $X = \{X_b\}_{b \in \mathbb{B}_L^d}$ : i.i.d., integrable)  $\Rightarrow$  SAW on random conductors.

Theorem 3 (Chino-Sakai (2016))

If the law of  $X$  is translation-invariant and ergodic, then the quenched critical point  $\hat{h}_X(x)$  is a.s. independent of the starting point  $x$  of SAWs.

- For SAW, percolation, OP, CP for  $d > d_c$ ,

$$p_c = 1 + CL^{-d} + O(L^{-d-1}), \quad C = \begin{cases} \sum_{n=2}^{\infty} U^{*n}(o) & [\text{SAW \& CP}], \\ \frac{1}{2} \sum_{n=2}^{\infty} U^{*2n}(o) & [\text{OP}], \\ U^{*2}(o) + \sum_{n=3}^{\infty} \frac{n+1}{2} U^{*n}(o) & [\text{percolation}]. \end{cases}$$

- For LT & LA for  $d > 8$ ,

$$p_c = \frac{1}{e} + CL^{-d} + O(L^{-d-1}), \quad C = \sum_{n=2}^{\infty} \frac{n+1}{2e} U^{*n}(o) + \begin{cases} 0 & [\text{LT}], \\ -\frac{1}{2e^2} \sum_{n=3}^{\infty} U^{*n}(o) & [\text{LA}]. \end{cases}$$

- The critical point is a divergent point of  $\chi_p = \sum_x \tau_p(x)$ . E.g., for SAW,  
**Lace expansion**  $\Rightarrow \tau_p = \delta_o + (pD + \pi_p) * \tau_p \Rightarrow \chi_p = (1 - p - \hat{\pi}_p)^{-1}$ .
- $p = e^{-h-X_b}$  ( $X = \{X_b\}_{b \in \mathbb{B}_L^d}$ : i.i.d., integrable)  $\Rightarrow$  SAW on random conductors.

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