THE DERRIDA–RETAUX CONJECTURE ON RECURSIVE MODELS

BY XINXING CHEN*, VICTOR DAGARD†, BERNARD DERRIDA†, YUEYUN HU‡, MIKHAIL LIFSHITS§, AND ZHAN SHI†

Shanghai Jiaotong University, École Normale Supérieure, Collège de France, Université Sorbonne Paris Nord, St. Petersburg State University, and Sorbonne Université Paris VI

We are interested in the nearly supercritical regime in a family of max-type recursive models studied by Collet, Eckman, Glaser and Martin [6] and by Derrida and Retaux [8], and prove that under a suitable integrability assumption on the initial distribution, the free energy vanishes at the transition with an essential singularity with exponent $\frac{1}{2}$. This gives a weaker answer to a conjecture of Derrida and Retaux [8]. Other behaviours are obtained when the integrability condition is not satisfied.

1. Introduction.

1.1. The model and main results. Let $m \geq 2$ be an integer. Let $X_0 \geq 0$ be a random variable taking values in $\mathbb{Z}_+ := \{0, 1, 2, \ldots\}$; to avoid triviality, it is assumed, throughout the paper, that $\mathbb{P}(X_0 \geq 2) > 0$. Consider the following recurrence relation: for all $n \geq 0$,

$$X_{n+1} = (X_{n,1} + \cdots + X_{n,m} - 1)^+,$$

where $X_{n,i}, i \geq 1$, are independent copies of $X_n$. Notation: $x^+ := \max\{x, 0\}$ for all $x \in \mathbb{R}$.

From (1.1), we get

$$m \mathbb{E}(X_n) - 1 \leq \mathbb{E}(X_{n+1}) \leq m \mathbb{E}(X_n),$$

which enables us to define the free energy

$$F_\infty := \lim_{n \to \infty} \downarrow \frac{\mathbb{E}(X_n)}{m^n} = \lim_{n \to \infty} \uparrow \frac{\mathbb{E}(X_n) - \frac{1}{m-1}}{m^n} \geq 0.$$

*Partially supported by NSFC grants 11771286 and 11531001.
†Partially supported by ANR project MALIN 16-CE93-0003.
‡Partially supported by ANR project SWIWS 17-CE40-0032-02.
§Partially supported by RFBR-DFG grant 20-51-12004.

MSC 2010 subject classifications: Primary 60J80; secondary 82B27

Keywords and phrases: Max-type recursive model, free energy.
We now recall a conjecture of Derrida and Retaux [8]. For any random variable $X$, we write $P_X$ for its law. Assume
\[ P_{X_0} = (1 - p) \delta_0 + p P_{X_0^*} , \]
where $\delta_0$ denotes the Dirac measure at 0, $X_0^*$ a (strictly) positive integer-valued random variable satisfying $P(X_0^* \geq 2) > 0$, and $p \in [0, 1]$ a parameter. Since $p \mapsto F_\infty =: F_\infty(p)$ is non-decreasing, there exists $p_c = p_c(X_0^*) \in [0, 1]$ such that $F_\infty(p) > 0$ for $p > p_c$ and that $F_\infty(p) = 0$ for $p < p_c$. The Derrida–Retaux conjecture says that if $p_c > 0$ (and possibly under some additional integrability conditions on $X_0$), then
\[ F_\infty(p) = \exp \left( - \frac{C + o(1)}{(p - p_c)^{1/2}} \right), \quad p \downarrow p_c , \quad (1.3) \]
for some constant $C \in (0, \infty)$. When $p_c = 0$, it is possible to have other exponents than $\frac{1}{2}$ in (1.3), see [15]. In [4], we have presented several open questions concerning the critical regime $p = p_c$ when $p_c > 0$.

The model with recursion defined in (1.1) is known to have a phase transition (Collet et al. [6]), recalled in Theorem A below. It is expected to have many universal properties at or near criticality, though few of these predicted properties have been rigorously proved so far. The model was introduced by Derrida and Retaux [8] as a simplified hierarchical renormalization model to understand the depinning transition of a line in presence of strong disorder [7]. The exponent $\frac{1}{2}$ in the Derrida–Retaux conjecture (1.3) was already predicted in [19], while another exponent 1 was predicted in [18]. For the mathematical literature of pinning models, see [2], [9], [10], [11], [12]. The same exponent 1 was found for a copolymer model in [3] with additional precision. The recursion (1.1) has appeared in another context, as a spin glass toy-model in Collet et al. [5]–[6], and is moreover connected to a parking scheme investigated by Goldschmidt and Przykucki [13]; it was studied from the point of view of iterations of random functions (Li and Rogers [17], Jordan [16]), and also figured as a special case in the family of max-type recursive models analyzed in the seminal paper of Aldous and Bandyopadhyay [1]. See Hu and Shi [15] for an extension to the case when $m$ is random, and Hu, Mallein and Pain [14] for an exactly solvable version in continuous time.

The aim of this paper is to study the Derrida–Retaux conjecture. Let us first recall the following characterisation of the critical regime.

---

\footnote{We are going to see that $p_c < 1$.}
Theorem A (Collet et al. [6]). We have

\begin{equation}
    p_c(X^*_0) = \frac{1}{1 + E\{(m-1)X^*_0 - 1|mX^*_0\}} \in [0,1].
\end{equation}

In words, assuming $E(X^*_0 mX^*_0) < \infty$, then $p = p_c$ means $E(mX^*_0) = (m-1)E(X_0 mX_0)$, and more precisely, $F_\infty > 0$ if $p > p_c$, while $F_\infty = 0$ if $p \leq p_c$.

It is natural to say that the system is subcritical if $p < p_c$, critical if $p = p_c$, and supercritical if $p > p_c$. Note from Theorem A that the assumption $p_c > 0$ in the Derrida–Retaux conjecture is equivalent to saying that $E(X^*_0 mX^*_0) < \infty$.

We give a partial answer to the Derrida–Retaux conjecture, by showing that under suitable general assumptions on the initial distribution, $\frac{1}{2}$ is the correct exponent, in the exponential scale, for the free energy.

**Theorem 1.1.** Assume $E[(X^*_0)^3 mX^*_0] < \infty$. Then

$$F_\infty(p) = \exp\left(-\frac{1}{(p - p_c)^{\frac{1}{2}+o(1)}}\right), \quad p \downarrow p_c.$$

It is possible to obtain some information about $o(1)$ in Theorem 1.1; see (3.1) and (7.1). A similar remark applies to Theorem 1.2 below.

It turns out that our argument in the proof of the lower bound (for the free energy) in Theorem 1.1 is quite robust. With some additional minor effort, it can be adapted to deal with systems that do not satisfy the condition $E[(X^*_0)^3 mX^*_0] < \infty$. Although this integrability condition might look exotic, it is optimal for the validity of the Derrida–Retaux conjecture. In the next theorem, we assume\(^2\) $P(X^*_0 = k) \sim c_0 m^{-k} k^{-\alpha}$, $k \to \infty$, for some constant $0 < c_0 < \infty$ and some parameter $2 < \alpha \leq 4$. The inequality $\alpha > 2$ ensures $p_c > 0$, which is the basic condition in the Derrida–Retaux conjecture, whereas the inequality $\alpha \leq 4$ implies $E[(X^*_0)^3 mX^*_0] = \infty$. It turns out that in this case, the behaviour of the free energy differs from the prediction in the Derrida–Retaux conjecture.

\(^2\)Notation: by $a_k \sim b_k$, $k \to \infty$, we mean $\lim_{k \to \infty} \frac{a_k}{b_k} = 1$. 
Theorem 1.2. Assume $\mathbf{P}(X_0^* = k) \sim c_0 m^{-k} k^{-\alpha}$, $k \to \infty$, for some $0 < c_0 < \infty$ and $2 < \alpha \leq 4$. Then

$$F_\infty(p) = \exp \left( -\frac{1}{(p - p_c)^{\nu + o(1)}} \right), \quad p \downarrow p_c,$$

where $\nu = \nu(\alpha) := \frac{1}{\alpha - 2}$.

Let us keep considering the situation $\mathbf{P}(X_0^* = k) \sim c_0 m^{-k} k^{-\alpha}$, $k \to \infty$, for some $0 < c_0 < \infty$. The case $2 < \alpha \leq 4$ was considered in Theorem 1.2. When $-\infty < \alpha \leq 2$, we have $\mathbf{E}(X_0^* m^{X_0^*}) = \infty$, which violates the basic condition $p_c > 0$ in the Derrida–Retaux conjecture; so the conjecture does not apply to this situation. In [15], it was proved that if $-\infty < \alpha < 2$, then $F_\infty(p) = \exp(-\frac{1}{\nu_1 + o(1)})$ when $p \downarrow p_c = 0$, with $\nu_1 = \nu_1(\alpha) = \frac{1}{2 - \alpha}$. This leaves us with the case $\alpha = 2$.

Theorem 1.3. Assume $\mathbf{P}(X_0^* = k) \sim c_0 m^{-k} k^{-2}$, $k \to \infty$, for some $0 < c_0 < \infty$. Then

$$F_\infty(p) = \exp \left( -\frac{e^{(C + o(1))}}{p} \right), \quad p \downarrow p_c = 0,$$

where $C := \frac{1}{(m - 1)c_0}$.

1.2. Description of the proof. Having in mind both the supercritical system (in Theorem 1.1) and the system with initial distribution satisfying $\mathbf{P}(X_0^* = k) \sim c_0 m^{-k} k^{-\alpha}$ with $2 < \alpha \leq 4$ or $\alpha = 2$ (in Theorems 1.2 and 1.3, respectively), we introduce in Section 2 a notion of regularity for systems. It is immediately seen that supercritical systems are regular (Lemma 2.4), and so are appropriately truncated systems in Theorems 1.2 and 1.3 (Lemma 8.4 in Section 8.2). Most of forthcoming technical results are formulated for regular systems in view of applications in the proof of Theorem 1.1 on the one hand, and of Theorems 1.2 and 1.3 on the other hand.

To study the free energy $F_\infty$, we make the simple observation that by (1.2), for all $n \geq 0$,

$$\frac{\mathbf{E}(X_n) - \frac{1}{m^{n-1}}}{m^n} \leq F_\infty \leq \frac{\mathbf{E}(X_n)}{m^n}, \quad n \geq 0.$$

So in order to bound $F_\infty$ from above, we only need to find a sufficiently large $n$ such that $\mathbf{E}(X_n) \leq 3$ (say), whereas to bound $F_\infty$ from below, it suffices to find another $n$ not too large, for which $\mathbf{E}(X_n) \geq 2$. 
Upper bound: the upper bound in the theorems is proved by studying the moment generating function. In the literature, the moment generating function is a commonly used tool to study the recursive system ([6], [8], [4]). In Section 3, we obtain a general upper bound (Proposition 3.1) for $E(X_n)$ for regular systems. Applying Proposition 3.1 to the supercritical system yields the upper bound in Theorem 1.1.

Lower bound: the proof of the lower bound in Theorem 1.1 requires some preparation. In Section 4, an elementary coupling, called the $XY$-coupling in Theorem 4.1, is presented for the supercritical system $(X_n)$ and a critical system $(Y_n)$, in such a way that $Y_n \leq X_n$ for all $n$. We then use a natural hierarchical representation of the systems, and study $N(0)_n$, the number of open paths (the paths on the genealogical tree along which the operation $x \mapsto x^+$ is unnecessary) up to generation $n$ with initial zero value. The most important result in Section 4 is the following inequality: if $E(X_0 - Y_0) \geq \eta P(Y_0 = 0)$ for some $\eta > 0$, then for suitable non-negative $r$, $n$, $k$, and $\ell$, (4.3) $E(X_{n+k+\ell}) \geq \frac{m^{k+\ell} \eta}{2} E\left[N_n^{(0)} 1_{\{N_n^{(0)} \geq r\}} 1_{\{Y_n = k\}}\right]$; see Theorem 4.2. This inequality serves as a bridge connecting, on the one hand, the expected value of the supercritical system $(X_n)$, and on the other hand, the expected number of open paths in the critical system $(Y_n)$.

Using (4.3) and the upper bound for $E(X_n)$ established in Proposition 3.1, we obtain an upper bound for $E\left[m Y_n \left(1 + Y_n\right) N(0)_n\right]$ for all $n$ and suitable $r = r(n)$; see Corollary 4.3. This application of (4.3) is referred to as the first crossing of the bridge, and is relatively effortless.

We intend to cross the bridge for a second time, but in the opposite direction. To prepare for the second crossing, we prove a recursive formula for $E[m Y_n (1 + Y_n) N(0)_n]$ in Proposition 5.1:

$$E[m Y_n (1 + Y_n) N(0)_n] = P(Y_0 = 0) \prod_{k=0}^{n-1} \left[E(m Y_k)\right]^{m-1}.$$  

Since the asymptotics of $\prod_{k=0}^{n-1} \left[E(m Y_k)\right]^{m-1}$ are known, this formula gives useful upper and lower bounds for $E[m Y_n (1 + Y_n) N(0)_n]$, stated in (5.15).

We are now ready to establish a good upper bound for $E(m Y_n N(0)_n)$ (see Lemma 6.2): on the event $\{N_n^{(0)} \geq r, Y_n \leq n\}$, the expectation was already handled by Corollary 4.3; on the complementary of this event (which needs to be split into two sub-cases), an application of the Markov inequality does the job thanks to the upper bound for $E[m Y_n (1 + Y_n) N(0)_n]$ in (5.15). [Actually
Lemma 6.2 states slightly less: it gives an upper bound only for the Cesàro sum of $E(mY_n N_n(0))$ — which nonetheless suffices for our needs.

The next step is to write a recursion formula for $E[(1 + Y_n)^3 mY_n N_n(0)]$ in the same spirit as Proposition 5.1; together with the upper bound for $E(mY_n N_n(0))$ in Lemma 6.2, the formula gives an upper bound for $E((1 + Y_n)^3 mY_n N_n(0))$; this is Proposition 6.3. [The power 3 in $(1 + Y_n)$ is important as we are going to see soon.]

Let us write, for positive integers $r$ and $K$,
\[
E[(1 + Y_n)mY_n N_n(0) 1_{\{N_n(0) \geq r\}} 1_{\{Y_n < K\}}] 
\geq E[(1 + Y_n)mY_n N_n(0)] - E[(1 + Y_n)mY_n N_n(0) 1_{\{Y_n \geq K\}}] 
- r E[(1 + Y_n)mY_n].
\]

We can bound $E[(1 + Y_n)mY_n N_n(0)]$ from below by means of Proposition 5.1 (or rather its consequence (5.15)), and bound $E[(1 + Y_n)mY_n N_n(0) 1_{\{Y_n \geq K\}}]$ from above by the Markov inequality and Proposition 6.3 (which is why the factor $(1 + Y_n)^3$ in the proposition is important, otherwise the bound would not be good enough), whereas $E[(1 + Y_n)mY_n]$ is smaller than a constant depending only on $m$. Consequently, we can choose appropriate values for $r$ and $K$ (both depending on $n$) and obtain a lower bound for $E[(1 + Y_n)mY_n N_n(0) 1_{\{N_n(0) \geq r\}} 1_{\{Y_n < K\}}]$. Since
\[
\max_{k \in [0, K) \cap \mathbb{Z}} (1 + k)m^k E[N_n(0) 1_{\{N_n(0) \geq r\}} 1_{\{Y_n = k\}}] 
\geq \frac{1}{K} E[(1 + Y_n)mY_n N_n(0) 1_{\{N_n(0) \geq r\}} 1_{\{Y_n < K\}}],
\]
there exists an integer $k \in [0, K)$ for which we have a lower bound for $E[N_n(0) 1_{\{N_n(0) \geq r\}} 1_{\{Y_n = k\}}]$. The parameters are chosen such that when using the bridge inequality (4.3) for the second time, we get $E(X_{n+k+\ell}) \geq 2$ for convenient $n$, $k$ and $\ell$. Together with the first inequality in (1.5), this yields the lower bound in Theorem 1.1.

Finally in Section 8, we prove Theorems 1.2 and 1.3 by means of a truncation argument.

The rest of the paper is as follows:
• Section 2: a notion of regularity for systems;
• Section 3: proof of the upper bound in Theorem 1.1;
• Section 4: XY-coupling, open paths, a lower bound via open paths;
• Section 5: a formula for the number of open paths and other preparatory work;
• Section 6: a general lower bound for free energy;
• Section 7: proof of the lower bound in Theorem 1.1;
• Section 8: proof of Theorems 1.2 and 1.3.

Notation: we often write $P_p$ instead of $P$ in order to stress dependence
on the parameter $p$ (accordingly, the corresponding expectation is denoted
by $E_p$), and $P_{p_c + \varepsilon}$ if $p = p_c + \varepsilon$. When we take $p = p_c + \varepsilon$, it is implicitly
assumed that $\varepsilon \in (0, 1 - p_c)$. Moreover, $\Lambda := 0$ for $\Lambda = \infty$ and $a \in \mathbb{R}$.

2. A notion of regularity for systems. Consider a generic $\mathbb{Z}_+\text{-valued}$
system $(X_n, n \geq 0)$ defined by (1.1), with $E(X_0^m X_0) < \infty$ (which is equiv-
alent to $p_c > 0$) and $P(X_0 \geq 2) > 0$.

In several situations we will assume that the system satisfies a certain
additional regularity condition, which is, actually, satisfied if the system is
critical or supercritical, or if it is suitably truncated.

**Definition 2.1.** Let $\zeta$ be a $\mathbb{Z}_+\text{-valued}$ random variable, with $E(\zeta^m \zeta) < \infty$ and $P(\zeta \geq 2) > 0$. Let $\beta \in [0, 2]$ and $\chi \in (0, 1]$. Write
\[
\Lambda(\zeta) := E(\zeta^3 \zeta) \in (0, \infty],
\]
\[
\Xi_k(\zeta) := E\left[\left(\zeta \wedge k\right)^2 \left[(m - 1)\zeta - 1\right] m^k\right] \in (0, \infty), \quad k \geq 1,
\]
where $a \wedge b := \min\{a, b\}$.

We say that the random variable $\zeta$ is $\beta$-regular with coefficient $\chi$ if for
all integers $k \geq 1$,
\[
\Xi_k(\zeta) \geq \chi \min\{\Lambda(\zeta), k^\beta\}.
\]

Furthermore, we say that a system $(X_n, n \geq 0)$ is $\beta$-regular with coeffi-
cient $\chi$, if $X_0$ is $\beta$-regular with coefficient $\chi$.

**Remark 2.2.** It is immediately seen that if the $\mathbb{Z}_+\text{-valued}$ random vari-
able $\zeta$ is 2-regular, then $\Lambda(\zeta) < \infty$ (otherwise, (2.3) would become: $\Xi_k(\zeta) \geq \chi k^2$ for all integers $k \geq 1$, which would lead to a contradiction, because
$\lim_{k \to \infty} \Xi_k(\zeta)/k^2 = 0$ by the dominated convergence theorem).  

**Remark 2.3.** Let us say a few words about our interest in the notion of the system’s regularity. Part of our concern is to obtain an upper bound for the moment generating function (the forthcoming Proposition 3.1, for
supercritical systems), which plays a crucial role in both upper and lower bounds in Theorems 1.1, 1.2 and 1.3. In this regard, the notion of regularity can be seen as a kind of (stochastic) lower bound for \( X_0 \) when the system is supercritical. The notion of regularity, however, is more frequently used in Sections 4–6 and 8, where it is applied to critical systems. These critical systems either have a finite value of the corresponding \( \Lambda(\cdot) \) (and they will be applied to prove Theorem 1.1), or are truncated critical systems whose values of \( \Lambda(\cdot) \) are finite but depending on the level of truncation (and will be applied to prove Theorems 1.2 and 1.3). In these cases, the notion of regularity should not be viewed as any kind of lower bound for (the initial distribution of) the system.

Recall that under the integrability condition \( E(X_0 m^{X_0}) < \infty, \), \( p \geq p_c \) means \( (m - 1)E(X_0 m^{X_0}) \geq E(m^{X_0}) \).

**Lemma 2.4.** If \( p \in [p_c, 1) \), then \((X_n, n \geq 0)\) is 0-regular with coefficient \( \chi = 1 - p \).

**Proof.** Let \( k \geq 1 \) be an integer. By definition,

\[
E_k(X_0) = E[(X_0 \wedge k)^2 \{(m - 1)X_0 - 1\} m^{X_0} 1_{\{X_0 \geq 1\}}].
\]

Since \((X_0 \wedge k)^2 \geq 1\) on \( \{X_0 \geq 1\}\), this yields

\[
E_k(X_0) \geq E[(m - 1)X_0 - 1] m^{X_0} 1_{\{X_0 \geq 1\}} = E[(m - 1)X_0 - 1] m^{X_0} + P(X_0 = 0).
\]

Recall that \( p \geq p_c \) means \( E\{(m - 1)X_0 - 1\} m^{X_0} \geq 0\), whereas \( P(X_0 = 0) = 1 - p\). This implies \( E_k(X_0) \geq 1 - p\), which is greater than or equal to \((1 - p)\min\{\Lambda(X_0), 1\}\). The lemma is proved.

**Remark 2.5.** Let \( \alpha \in [2, 4] \). Assume \( P(X_0^* = k) \sim c_0 m^{-k-\alpha}, k \to \infty\), for some \( 0 < c_0 < \infty \). We are going to see in Lemma 8.4 that a critical system started at a conveniently truncated version of \( X_0^* \) is \((4 - \alpha)\)-regular. This will allow to use a truncation argument in Section 8 to prove Theorems 1.2 and 1.3.
3. Proof of Theorem 1.1: upper bound. The upper bound in Theorem 1.1 does not require the full assumption in the theorem. In particular, here we do not need the assumption $\mathbb{E}(X_0^3 m^{X_0}) < \infty$.

Throughout the section, we assume $p > p_c > 0$, i.e., $\mathbb{E}(m^{X_0}) < (m-1)\mathbb{E}(X_0 m^{X_0}) < \infty$. The upper bound in Theorem 1.1 is as follows: there exists a constant $c_1 > 0$ such that for all sufficiently small $\varepsilon > 0$,

\[
F_\infty(p_c + \varepsilon) \leq \exp \left( -\frac{c_1}{\varepsilon^{1/2}} \right).
\]

Let

$$\Lambda = \Lambda(X_0) := \mathbb{E}(X_0^3 m^{X_0}) \in (0, \infty],$$

as in (2.1). When $p \geq p_c$, we have $\mathbb{E}(m^{X_0}) \leq (m-1)\mathbb{E}(X_0 m^{X_0})$ by definition, so $\Lambda \geq \mathbb{E}(X_0 m^{X_0}) \geq \frac{1}{m-1} \mathbb{E}(m^{X_0}) \geq \frac{1}{m-1}$. Consequently,

\[
\Lambda \in \left[ \frac{1}{m-1}, \infty \right] \text{ if } p \geq p_c.
\]

The main step in the proof of (3.1) is the following proposition. We recall from Remark 2.2 that if a system is 2-regular in the sense of (2.3), then $\Lambda < \infty$.

**Proposition 3.1.** Let $\beta \in [0, 2]$ and $\chi \in (0, 1]$. Assume $\mathbb{E}(X_0 m^{X_0}) < \infty$ and the system $(X_n, n \geq 0)$ is $\beta$-regular with coefficient $\chi$ in the sense of (2.3). Let $\delta_0 := (m-1)\mathbb{E}(X_0 m^{X_0}) - \mathbb{E}(m^{X_0})$. There exist constants $c_2 \in (0, 1]$ and $c_3 \in (0, 1]$, depending only on $m$ and $\beta$, such that if $\delta_0 \in (0, c_3)$ (for $\beta \in [0, 2)$) or if $\delta_0 \in (0, c_3 \chi)$ (for $\beta = 2$), then there exists an integer $n_0 \geq K_0$ satisfying

\[
\mathbb{E}(m^{X_{n_0}}) \leq 3, \quad \max_{0 \leq i \leq n_0} \mathbb{E}(X_i) \leq 3,
\]

where $K_0 := \min\{(\frac{c_2 \Lambda \chi}{\delta_0})^{1/2}, (\frac{c_3 \Lambda \chi}{\delta_0})^{1/2-\beta}\}$ for $\beta \in [0, 2)$ and $K_0 := (\frac{c_2 \Lambda \chi}{\delta_0})^{1/2}$ for $\beta = 2$.

By admitting Proposition 3.1 for the time being, we are able to prove the upper bound (3.1) in Theorem 1.1.

**Proof of Theorem 1.1: upper bound.** Recall from (1.5) that for all $n \geq 0$,

$$\frac{\mathbb{E}(X_n) - \frac{1}{m-1} m^n}{m^n} \leq F_\infty \leq \frac{\mathbb{E}(X_n)}{m^n}, \quad n \geq 0.$$
Lemma 2.4 says that when \( p \in (p_c, 1) \), the system \((X_n, n \geq 0)\) is 0-regular with coefficient \( \chi = 1 - p \), so we are entitled to apply Proposition 3.1 to \( \beta = 0 \). We take \( p = p_c + \varepsilon \) (with \( 0 < \varepsilon < 1 - p_c \)); note that

\[
\delta_0 := (m - 1)E_{p_c+\varepsilon}(X_0^mX^0) - E_{p_c}(mX^0) = \left( (m - 1)E_{p_c}(X_0^mX^0) - E_{p_c}(mX^0) \right) + \varepsilon \left( E\{(m - 1)X^*_0 - 1|mX^0\} + 1 \right).
\]

By Theorem A in the introduction, \((m - 1)E_{p_c}(X_0^mX^0) - E_{p_c}(mX^0) = 0\). So \( \delta_0 = c_4 \varepsilon \) where \( c_4 := E\{(m - 1)X^*_0 - 1|mX^0\} + 1 \in (1, \infty) \). By Proposition 3.1 (with \( \beta = 0 \)), \( E_{p_c+\varepsilon}(X_{[c_5/\varepsilon^{1/2}]}^mX^0) \leq 3 \) for some constant \( c_5 > 0 \) and all sufficiently small \( \varepsilon > 0 \). Using the second inequality in (1.5), we obtain the following bound for the free energy: for all sufficiently small \( \varepsilon > 0 \),

\[
F_\infty(p_c + \varepsilon) \leq \frac{E_{p_c+\varepsilon}(X_{[c_5/\varepsilon^{1/2}]}^mX^0)}{m_{[c_5/\varepsilon^{1/2}]}^m} \leq 3 \exp \left( - \frac{c_5}{\varepsilon^{1/2}} - 1 \right) \log m,
\]

which yields (3.1).

The rest of the section is devoted to the proof of Proposition 3.1. Assume \( E(X^*_0^mX^0) < \infty \) (which is equivalent to saying that \( p_c > 0 \)) and \( p > p_c \). For all \( n \geq 0 \), we write the moment generating function

\[ H_n(s) := E(s^{X^n}). \]

We rewrite the iteration equation (1.1) in terms of \( H_n \): for all \( n \geq 0 \),

\[
H_{n+1}(s) = \frac{1}{s} H_n(s)^m + (1 - \frac{1}{s}) H_n(0)^m.
\]

A useful quantity in the proof is, for \( n \geq 0 \),

\[ \delta_n := m(m - 1)H'_n(m) - H_n(m) = (m - 1)E(X_n^mX^0) - E(mX^0). \]

By assumption, \( \delta_0 > 0 \) (and is small). Using the iteration relation (3.4), it is immediate that

\[ \delta_{n+1} = H_n(m)^{m-1} \delta_n. \]

Consequently, for \( n \geq 1 \),

\[
\delta_n = \left( \prod_{i=0}^{n-1} H_i(m)^{m-1} \right) \delta_0 \in (0, \infty).
\]

[The recursion formula was known to Collet et al. [6]; see also Equation (10) in [4].] We outline the proof of Proposition 3.1 before getting into details.
Outline of the proof of Proposition 3.1. Only the first inequality (saying that $E(m^{n_0}) \leq 3$ for some integer $n_0 \geq K_0$) in the proposition needs to be proved. Let $\theta > 0$ be a constant. We choose

$$n_0 := \sup \{ i \geq 1 : \delta_i \leq \theta \}.$$  

[If $n_0 = \infty$, the proposition is easily proved. So we assume $n_0 < \infty$.] Since it is quite easy to see that $H_k(m) \leq m^{1/(m-1)}e^{\delta_i}$ for $i \geq 0$, we get $H_{n_0}(m) \leq m^{1/(m-1)}e^{\theta} \leq 3$ if $\theta$ is chosen to satisfy $m^{1/(m-1)}e^{\theta} \leq 3$. Consequently, $E(m^{n_0}) \leq 3$.

It remains to check that $n_0 \geq K_0$. The key ingredient is the following inequality: for all $0 \leq n \leq n_0$,

$$\frac{\delta_n}{\delta_0} = \prod_{i=0}^{n-1} H_i(m)^{m-1} \leq \frac{c_6}{\Theta_0(s_n)},$$

where $c_6 > 0$ is a constant depending only on $m$, $s_n := m(1 - \frac{C_1}{n+1})$ for some constant $C_1 \in (0, 1)$ depending on $m$, and $\Theta_i(\cdot)$, for $i \geq 0$, is a positive function on $(0, m)$ which is to be defined soon. [It turns out that for $s \in (0, m)$, $\Theta_0(s)$ is connected to $\Xi_k(X_0)$ defined in (2.2), with $k = k(s) := \lfloor \frac{m}{s-2} \rfloor$. This helps explain partly the importance of the notion of regularity for systems.]

Let us look at (3.16) with $n := n_0$. We have $\delta_{n_0} = \frac{\delta_{1+n_0}}{\prod_{i=0}^{n_0} H_i(m)^{m-1}} > \frac{\theta}{\prod_{i=0}^{n_0} H_i(m)^{m-1}}$. Since $H_{n_0}(m) \leq 3$ as we have already seen, it follows that $\delta_{n_0}$ is greater than the positive constant $\frac{\theta}{3m-1}$. Using a simple lower bound for $\Theta_0(s_n)$ (thanks to the aforementioned connection between $\Theta_0(s)$ and $\Xi_k(X_0)$) as a function of $n$, (3.16) will yield $n_0 \geq K_0$ as desired.

We are thus left with the proof of (3.16). The idea is to study not only the function $\Theta_0(\cdot)$, but the sequence of functions $\Theta_n(\cdot)$, $n \geq 0$, and produce a recursion in $n$: for $0 \leq n \leq n_0$, $s \in \left[ \frac{m}{2}, m \right)$ and some constant $C_2 > 1$

$$\Theta_{n+1}(s) \geq [C_2 s - (C_2 - 1)m] \frac{H_n(s)^{m-1}}{s} \Theta_n(s).$$

The proof of this inequality, done in two steps (Lemmas 3.3 and 3.4), follows the lines of [4] and [6].

It is quite easy to see that $\frac{H_n(s)^{m-1}}{s} \geq (1 - C_3(m - s)) \frac{H_n(m)^{m-1}}{m}$ for some constant $C_3 > 0$. So, as long as $s \in \left[ \frac{m}{2}, m \right)$ satisfies $C_2 s - (C_2 - 1)m > 0$ and

\[\prod_{i=0}^{n_0} \Theta_i := 1.\]
1 - C_3(m - s) > 0 (which is the case when we choose \( s := s_n = m(1 - \frac{C_1}{n+1}) \) later), we get

\[
\Theta_{n+1}(s) \geq \left[ \frac{C_2 s - (C_2 - 1)m}{m} (1 - C_3(m - s)) \right] \frac{H_n(m)^{m-1} \Theta_n(s)}{H_n(s)},
\]

and thus, by iteration,

\[
\Theta_n(s) \geq \left[ \frac{C_2 s - (C_2 - 1)m}{m} (1 - C_3(m - s)) \right]^n \Theta_0(s) \prod_{i=0}^{n-1} H_i(m)^{m-1}.
\]

On the other hand, \( \Theta_n(s) \leq 2 \) for \( 0 < s < m \) and \( 0 \leq n \leq n_0 \). With our choice of \( s = s_n \), the factor \( \left[ \frac{C_2 s - (C_2 - 1)m}{m} (1 - C_3(m - s)) \right]^n \) is greater than a positive constant independent of \( n \). This will imply (3.16). \( \square \)

We now proceed to the detailed proof of Proposition 3.1. Let us start with a few elementary properties of the moment generating functions.

**Lemma 3.2.** Assume \( \mathbb{E}(X_0^* m^{X_0^*}) < \infty \) and \( p > p_c \). Let \( n \geq 0 \).

(i) The functions \( s \mapsto (m - 1)sH'_n(s) - H_n(s) \) and \( s \mapsto (\frac{m-1}{s})H'_n(s) - H_n(s) \) are non-decreasing on \((0, m]\).

(ii) We have \( H_n(m) \leq m^{1/(m-1)}e^{\delta_n} \).

(iii) For \( s \in [1, m] \),

\[
\frac{H_n(s)^{m-1}}{s} \geq \frac{H_n(m)^{m-1}}{m} [1 - m(m-s) \delta_n e^{(m-1)\delta_n}].
\]

**Proof.** (i) Write

\[
\varphi_n(s) := (m - 1)sH'_n(s) - H_n(s).
\]

[So \( \varphi_n(m) = \delta_n \).] For all \( s \in (0, m), \)

\[
\varphi'_n(s) = (m - 2)H'_n(s) + (m - 1)sH''_n(s) \geq 0,
\]

which implies the monotonicity of \( \varphi_n \). In particular,

\[
\varphi_n(s) \leq \delta_n, \quad s \in (0, m).
\]

To prove the monotonicity of \( s \mapsto \frac{\varphi_n(s)}{s} \), we note that

\[
\frac{d}{ds} \left( \frac{\varphi_n(s)}{s} \right) = \frac{s \varphi'_n(s) - \varphi_n(s)}{s^2} = \frac{(m - 1)s^2 H''_n(s) + H_n(s) - sH'_n(s)}{s^2}.
\]
On the right-hand side, the numerator is greater than or equal to $s^2 H''_n(s) + H'_n(s) - s H'_n(s)$, which is $\mathbb{E}\{(X_n(X_n - 1) + 1 - X_n)s^{X_n}\} = \mathbb{E}\{(X_n - 1)^2 s^{X_n}\} \geq 0$, so the desired result follows.

(ii) For $s \in [1, m)$,

$$
\frac{d}{ds} \left( \frac{H_n(s)^{m-1}}{s} \right) = \left[ (m - 1) s H'_n(s) - H_n(s) \right] \frac{H_n(s)^{m-2}}{s^2}.
$$

By (i), $(m - 1) s H'_n(s) - H_n(s) \leq (m - 1) m H'_n(m) - H_n(m) = \delta_n$. On the other hand, $\frac{H_n(s)^{m-2}}{s^2} \leq \frac{H_n(s)^{m-1}}{s}$ (because $s \geq 1$ and $H_n(s) \geq 1$). Hence

$$
\frac{d}{ds} \left( \frac{H_n(s)^{m-1}}{s} \right) \leq \delta_n \frac{H_n(s)^{m-1}}{s}, \quad s \in [1, m).
$$

When $s = 1$, $\frac{H_n(s)^{m-1}}{s}$ equals 1, so solving the differential equation yields that for $s \in [1, m)$,

$$
\frac{H_n(s)^{m-1}}{s} \leq e^{\delta_n(s-1)}.
$$

Taking $s = m$ yields the desired inequality.

For further use, we also observe that our proof yields the following inequality in the critical case: if $(m - 1) \mathbb{E}(Y_0 m^Y_0) = \mathbb{E}(m^Y_0) < \infty$, then

$$
\mathbb{E}(m^Y_0) \leq m^{1/(m-1)}, \quad n \geq 0.
$$

(iii) For $s \in [1, m)$, we have, by (3.9) and (3.10), $\frac{d}{ds} \left( \frac{H_n(s)^{m-1}}{s} \right) \leq \delta_n e^{(m-1)\delta_n}$. So for $s \in [1, m]$,

$$
\frac{H_n(s)^{m-1}}{s} \geq \frac{H_n(m)^{m-1}}{m} - (m - s) \delta_n e^{(m-1)\delta_n}.
$$

Since $H_n(m)^{m-1} \geq 1$, we have $(m - s) \delta_n e^{(m-1)\delta_n} \leq \frac{H_n(m)^{m-1}}{m} m (m - s) \delta_n e^{(m-1)\delta_n}$, implying the desired conclusion. \hfill \Box

Define

$$
\Theta_n(s) := [H_n(s) - s(s - 1) H''_n(s)] - \frac{(m - 1)(m - s)}{m} [2s H'_n(s) + s^2 H''_n(s)] + \delta_n.
$$

[In the critical regime, $\delta_n = 0$ for all $n$, so $\Theta_n(\cdot)$ is the function $\Delta_n(\cdot)$ studied in [4].] Following the lines of [4] and [6], we first prove two preliminary inequalities for $\Theta_n$. 

Lemma 3.3. Assume \( \mathbb{E}(X_0^m m^{X_0^m}) < \infty \), and \( p > p_c \). Let \( n \geq 0 \).

(i) We have \( \Theta_n(s) \in [0, 1 + \delta_n] \) for all \( s \in [0, m] \).

(ii) If \( \delta_n \leq 1 \), then for all \( s \in \left[ \frac{m}{2}, m \right) \),

\[
\Theta_{n+1}(s) \geq \frac{m}{s} \Theta_n(s) H_n(s)^{m-1} - \frac{\kappa(m-s)}{s} [\delta_n - \varphi_n(s)]^2 H_n(s)^{m-2},
\]

where \( \kappa := 3^{m-2} + 1 \), and \( \varphi_n(s) := (m-1)sH_n'(s) - H_n(s) \) as in (3.6).

Proof. (i) By definition, for \( s \in (0, m) \),

\[
\Theta'_n(s) = -\frac{m-s}{m} [2(m-2)H_n'(s) + (4m-5)H_n''(s) + (m-1)s^2H_n'''(s)] \leq 0,
\]

so \( s \mapsto \Theta_n(s) \) is non-increasing on \((0, m)\). Since \( \lim_{s \to m^-} \Theta_n(s) = 0 \), and \( \Theta_n(0) = H_n(0) + \delta_n \leq 1 + \delta_n \), the result follows.

(ii) Let \( s \in \left[ \frac{m}{2}, m \right) \). The iteration (3.4) yields

\[
\Theta_{n+1}(s) = \frac{m}{s} \Theta_n(s) H_n(s)^{m-1} - \frac{m-s}{s} \varphi_n(s)^2 H_n(s)^{m-2} + m \left( \frac{H_n(m)^{m-1}}{m} - \frac{H_n(s)^{m-1}}{s} \right) \delta_n.
\]

(3.12)

For the last term on the right-hand side, we recall from (3.8) that \( \frac{d}{ds} \left( \frac{H_n(s)^{m-1}}{s} \right) = \varphi_n(s) \frac{H_n(s)^{m-2}}{s^2} \), so by the mean-value theorem, there exists \( y \in [s, m] \) such that

\[
\frac{H_n(m)^{m-1}}{m} - \frac{H_n(s)^{m-1}}{s} = (m-s) \frac{\varphi_n(y)}{y} \frac{H_n(y)^{m-2}}{y} \geq (m-s) \frac{\varphi_n(s)}{s} \frac{H_n(y)^{m-2}}{y},
\]

the last inequality being a consequence of Lemma 3.2 (i). Going back to (3.12), we see that the proof will be finished if we are able to check that for all \( u \in [s, m] \), with \( \kappa := 3^{m-2} + 1 \),

\[
\varphi_n(s)^2 H_n(s)^{m-2} - \frac{m}{u} \varphi_n(s) H_n(u)^{m-2} \delta_n \leq \kappa[\delta_n - \varphi_n(s)]^2 H_n(s)^{m-2}.
\]

(3.13)

We prove (3.13) by distinguishing two possible situations. We write \( \text{LHS}_{(3.13)} \) for the expression on the left-hand side of (3.13), and \( \text{RHS}_{(3.13)} \) for the expression on the right-hand side.

First situation: \( \varphi_n(s) \geq 0 \). We use the trivial inequalities \( \frac{m}{u} \geq 1 \) and \( H_n(u)^{m-2} \geq H_n(s)^{m-2} \), to see that

\[
\text{LHS}_{(3.13)} \leq \varphi_n(s)^2 H_n(s)^{m-2} - \varphi_n(s) H_n(s)^{m-2} \delta_n,
\]
which is non-positive (because \( \varphi_n(s) \leq \delta_n \); see (3.7)). This yields (3.13) since \( \text{RHS}_{(3.13)} \geq 0 \).

Second (and last) situation: \( \varphi_n(s) < 0 \). We write \(|\varphi_n(s)|\) instead of \(-\varphi_n(s)\) in this situation. We have

\[
H_n(m) \leq H_n(s) + (m - s)H'_n(m) = H_n(s) + (m - s)\frac{H_n(m) + \delta_n}{m(m - 1)}.
\]

For \( s \in \left[\frac{m}{2}, m\right) \), we have \( \frac{m-s}{m(m-1)} \leq \frac{m-s}{m} \leq \frac{1}{2} \), so

\[
H_n(m) \leq H_n(s) + \frac{H_n(m) + \delta_n}{2}.
\]

Consequently,

\begin{equation}
(3.14) \quad H_n(m) \leq 2H_n(s) + \delta_n \leq 2H_n(s) + 1 \leq 3H_n(s),
\end{equation}

where we used the assumption \( \delta_n \leq 1 \) in the second inequality, and the trivial relation \( H_n(s) \geq 1 \) in the last inequality.

We now look at the second expression on the left-hand side of (3.13). The factor \( \frac{m}{u} \) is easy to deal with: we have \( \frac{m}{u} \leq 2 \) (using \( u \geq s \geq \frac{m}{2} \)). For the factor \( H_n(u)^{m-2} \), since \( u \leq m \), we have \( H_n(u) \leq H_n(m) \leq 3H_n(s) \) (by means of (3.14)), so \( H_n(u)^{m-2} \leq 3^{m-2}H_n(s)^{m-2} \). Consequently,

\[
\text{LHS}_{(3.13)} \leq |\varphi_n(s)|^2H_n(s)^{m-2} + 2|\varphi_n(s)|3^{m-2}H_n(s)^{m-2} \delta_n.
\]

We look at the two terms on the right-hand side. For the first term, we argue that \(|\varphi_n(s)|^2 \leq (|\varphi_n(s)| + \delta_n)^2 \). For the second term, we use \( 2|\varphi_n(s)|\delta_n \leq (|\varphi_n(s)| + \delta_n)^2 \). Hence

\[
\text{LHS}_{(3.13)} \leq (|\varphi_n(s)| + \delta_n)^2H_n(s)^{m-2} + (|\varphi_n(s)| + \delta_n)^23^{m-2}H_n(s)^{m-2}
= (3^{m-2} + 1)(|\varphi_n(s)| + \delta_n)^2H_n(s)^{m-2},
\]

which yields (3.13) again, as \([\delta_n - \varphi_n(s)]^2 = (|\varphi_n(s)| + \delta_n)^2 \) in this case. [Note that this case is very easy to handle when \( m = 2 \); all we need is to observe that \( \frac{m}{u} \leq 2 \).] \( \square \)

**Lemma 3.4.** Assume \( \mathbb{E}(X_0 m^{X_0}) < \infty \), and \( p > p_c \). Let \( n \geq 0 \). Then

\[
[\delta_n - \varphi_n(s)]^2 \leq 2(H_n(0) + \delta_n)\Theta_n(s), \quad s \in (0, m),
\]

where \( \varphi_n(s) := (m-1)sH'_n(s) - H_n(s) \) as in (3.6).
Proof. The lemma in the critical regime was already proved in ([4], proof of Lemma 7). The argument remains valid in our situation if we replace $G_n(s)$ and $\Delta_n(s)$ there (notation of [4]) by $H_n(s) + \delta_n$ and $\Theta_n(s)$, respectively. It is reproduced here for the sakes of clarity and self-containedness.

By definition, with $t := \frac{s}{m} \in (0, 1)$,

\[
\Theta_n(s) = [H_n(s) - s(s - 1)H'_n(s)] - [H_n(m) - m(m - 1)H'_n(m)] - \frac{(m - 1)(m - s)}{m} [2sH'_n(s) + s^2H''_n(s)]
\]

\[
= \mathbb{E}\left\{m^{X_n}[(m - 1)X_n - 1][1 - (1 + X_n)t^{X_n} + X_n t^{1+X_n}]\right\}
\]

(3.15) \[
= \mathbb{E}\left\{m^{X_n}[(m - 1)X_n - 1][1 - (1 + X_n)t^{X_n} + X_n t^{1+X_n}] \mathbf{1}_{\{X_n \geq 1\}}\right\},
\]

and

\[
\delta_n - \varphi_n(s) = \mathbb{E}\left\{m^{X_n}(1 - t^{X_n})[(m - 1)X_n - 1] \mathbf{1}_{\{X_n \geq 1\}}\right\},
\]

\[
H_n(0) + \delta_n = \mathbb{E}\left\{m^{X_n}[(m - 1)X_n - 1] \mathbf{1}_{\{X_n \geq 1\}}\right\}.
\]

By the Cauchy–Schwarz inequality,

\[
[\delta_n - \varphi_n(s)]^2 \leq (H_n(0) + \delta_n) \mathbb{E}\left\{m^{X_n}(1 - t^{X_n})^2[(m - 1)X_n - 1] \mathbf{1}_{\{X_n \geq 1\}}\right\}.
\]

So the proof of the lemma is reduced to showing the following: for $t \in (0, 1)$ and integer $k \geq 1$,

\[
(1 - t^k)^2 \leq 2[1 - (1 + k)t^k + kt^{1+k}].
\]

This is equivalent to saying that $2k t^k(1 - t) \leq 1 - t^{2k}$, which is obviously true because $t^{1+k} = 1 + t + \ldots + t^{2k-1} \geq 2kt^k$ (noting that $2t^k \leq 2t^{k-1/2} \leq t^{k+\ell} + t^{k-\ell-1}$ for $0 \leq \ell < k$).

We have now all the ingredients for the proof of Proposition 3.1.

Proof of Proposition 3.1. Assume \(\delta_0 := (m - 1)\mathbb{E}(X_0m^{X_0}) - \mathbb{E}(m^{X_0}) > 0\). If we are able to prove the first inequality (saying that \(\mathbb{E}(m^{X_{n_0}}) \leq 3\) for some integer \(n_0 \geq K_0\)), then \(\mathbb{E}(X_{n_0}) \leq \mathbb{E}(m^{X_{n_0}}) \leq 3\). Since \(m\mathbb{E}(X_n) - 1 \leq \mathbb{E}(X_{n+1})\), i.e., \(\mathbb{E}(X_n) \leq \frac{1 + \mathbb{E}(X_{n+1})}{m}\) (for all \(n \geq 0\)), the second inequality in Proposition 3.1 follows immediately.

It remains to prove the first inequality. Fix a constant \(0 < \theta \leq 1\) such that \(m^{1/(m-1)} \leq e^\theta \leq 3\). Let us define

\[
n_0 = n_0(\theta, \delta_0) := \sup\{i \geq 0 : \delta_i \leq \theta\}.
\]
We may assume that $\delta_0 \in (0, \frac{\theta}{m})$, so by Lemma 3.2 (ii), $\delta_1 = H_0(m)^{m-1} \delta_0 \leq [m^{1/(m-1)} e^{\delta_0}]^{m-1} \delta_0 < \theta$. This implies $n_0 \geq 1$.

If $n_0 = \infty$, then $\delta_i \leq \theta$ for all $i \geq 0$, and by Lemma 3.2 (ii), $E(m^{X_n}) \leq m^{1/(m-1)} e^{\delta_0} \leq m^{1/(m-1)} e^{\theta} \leq 3$ for all $n \geq 0$; there is nothing to prove in the proposition. In the following, we assume that $n_0 < \infty$.

Let $0 \leq n \leq n_0$ (hence $\delta_n \leq \theta \leq 1$) and $s \in (\frac{4\kappa-1}{4\kappa}m, m)$, where $\kappa := 3^{m-2} + 1$ as in Lemma 3.3. We reproduce the conclusion of Lemma 3.3 (ii) in its notation:

$$\Theta_n(s) \geq \frac{m}{s} \Theta_n(s) H_n(s)^{m-1} - \frac{\kappa (m-s)}{s} [\delta_n - \varphi_n(s)]^2 H_n(s)^{m-2}.$$

By Lemma 3.4, $[\delta_n - \varphi_n(s)]^2 \leq 2(H_n(0) + \delta_n) \Theta_n(s)$; since $H_n(0) \leq 1$ and $\delta_n \leq \theta \leq 1$ (for $0 \leq n \leq n_0$), we have $[\delta_n - \varphi_n(s)]^2 \leq 4\Theta_n(s) \leq 4\Theta_n(s) H_n(s)$ (because $H_n(s) \geq 1$ for $s \geq 1$); this implies that for $0 \leq n \leq n_0$,

$$\Theta_n(s) \geq \frac{m}{s} \Theta_n(s) H_n(s)^{m-1} - \frac{4\kappa (m-s)}{s} \Theta_n(s) H_n(s)^{m-1}$$

Iterating the procedure, we see that for $0 \leq n \leq n_0$ and $s \in (\frac{4\kappa-1}{4\kappa}m, m)$,

$$\Theta_n(s) \geq \Theta_0(s) [4\kappa s - (4\kappa - 1)m] \prod_{i=0}^{n-1} \frac{H_i(s)^{m-1}}{s}.$$

By Lemma 3.2 (iii),

$$\frac{H_n(s)^{m-1}}{s} \geq \frac{H_n(m)^{m-1}}{m} \left[1 - m(m-s) \delta_n e^{(m-1)\theta_n}\right],$$

which is greater than or equal to $\frac{H_n(m)^{m-1}}{m} \left[1 - m(m-s) \theta e^{(m-1)\theta}\right]$ for $0 \leq n \leq n_0$. Hence, for $0 \leq n \leq n_0$ and $s \in (\frac{4\kappa-1}{4\kappa}m, m)$ satisfying $m(m-s) \theta e^{(m-1)\theta} < 1$,

$$\prod_{i=0}^{n-1} H_i(m)^{m-1} \leq \frac{\Theta_n(s)}{\Theta_0(s)} \frac{m^n}{[4\kappa s - (4\kappa - 1)m]^n} \left[1 - m(m-s) \theta e^{(m-1)\theta}\right]^n.$$

We take $s = s_n := (1 - \frac{1}{8m^2\kappa(n+1)})m \in (\frac{4\kappa-1}{4\kappa}m, m)$ from now on. The requirement $m(m-s_n) \theta e^{(m-1)\theta} < 1$ is met with our choice: since $\theta \leq 1$ and $e^{\theta} \leq 3$, we have

$$m(m-s_n) \theta e^{(m-1)\theta} = \frac{\theta e^{(m-1)\theta}}{8\kappa(n+1)} \leq \frac{3^{m-1}}{8(3^{m-2}+1)(n+1)} < 1.$$
By Lemma 3.3 (i), $\Theta_n(s_n) \leq 1 + \delta_n \leq 1 + \theta \leq 2$ (for $0 \leq n \leq n_0$). This yields the existence of a constant $c_6 > 0$, depending only on $m$, such that for $0 \leq n \leq n_0$,

$$\prod_{i=0}^{n-1} H_i(m)^{m-1} \leq \frac{c_6}{\Theta_0(s_n)}.$$  

(3.16)

Let us have a closer look at $\Theta_0(s_n)$. Recall from (3.15) that for $s \in (1, m)$ and $t := \frac{s}{m} \in (0, 1)$,

$$\Theta_0(s) = E\left\{ m^{X_0} [(m - 1)X_0 - 1] [1 - (1 + X_0)t^{X_0} + X_0 t^{1 + X_0}] 1_{\{X_0 \geq 1\}} \right\}.$$  

For $t \in (0, 1)$ and $k \geq 1$, we have $t^k \leq e^{(1-t)k}$, so $1 - (1 + k)t^k + kt^{1+k} \geq 1 - (1 + u)e^{-u}$, where $u := (1 - t)k > 0$. Observe that $1 - (1 + v)e^{-v} \geq 1 - \frac{v^2}{2}$ for $v \geq 1$ (because $v \mapsto 1 - (1 + v)e^{-v}$ is increasing on $(0, \infty)$) and that $1 - (1 + v)e^{-v} \geq \frac{v^2}{2e}$ for $v \in (0, 1]$ (because $v \mapsto 1 - (1 + v)e^{-v} - \frac{v^2}{2e}$ is increasing on $(0, 1]$).

Hence

$$1 - (1 + k)t^k + kt^{1+k} \geq c_7 \min\{ (1 - t)^2k^2, 1 \}, \quad t \in (0, 1), \quad k \geq 1,$$

with $c_7 := \min\{1 - \frac{2}{e}, \frac{1}{2e}\} > 0$. Consequently, for all $s \in (0, m)$,

$$\Theta_0(s) \geq c_7 E\left\{ m^{X_0} [(m - 1)X_0 - 1] \min\{ (1 - \frac{s}{m})^2 X_0^2, 1 \} 1_{\{X_0 \geq 1\}} \right\} = c_7 (1 - \frac{s}{m})^2 E\left\{ m^{X_0} [(m - 1)X_0 - 1] \min\{ X_0^2, \frac{m^2}{(m - s)^2} \} \right\}.$$  

Let $\Xi_k(X_0) := E\{(X_0 \wedge k)^2 [(m - 1)X_0 - 1] m^{X_0}\}$ be as in (2.2). Then this gives

$$\Theta_0(s) \geq c_7 (1 - \frac{s}{m})^2 \Xi_{\frac{m}{m-s}}(X_0).$$

We now come back to our choice of $s = s_n := (1 - \frac{1}{8m^2\kappa(n+1)})m$, so $1 - \frac{s}{m} = \frac{1}{8m^2\kappa(n+1)}$, and $\frac{m-s}{m-s} = 8m^2\kappa(n+1)$. By assumption, the system $(X_n, n \geq 0)$ is $\beta$-regular with coefficient $\chi$ in the sense of (2.3), i.e., $\Xi_k(X_0) \geq \chi \min\{ \Lambda, k^\beta \}$ for all integers $k \geq 1$. Consequently, for $n \geq 0$, with $c_8 := \frac{c_7}{8m^2\kappa^2}$,

$$\Theta_0(s_n) \geq \frac{c_8 \chi}{(n+1)^2} \min\{ \Lambda, (n+1)^\beta \}.$$  

Combining with (3.16), we get, for $0 \leq n \leq n_0$ and with $c_9 := \frac{c_8}{c_6}$,

$$\prod_{i=0}^{n-1} H_i(m)^{m-1} \leq \frac{c_9}{\chi} \frac{(n+1)^2}{\min\{ \Lambda, (n+1)^\beta \}}.$$
Recall from (3.5) that \(\delta_n = (\prod_{i=0}^{n-1} H_i(m)^{m-1})\delta_0\), which yields: for \(0 \leq n \leq n_0\),
\[
\delta_n \leq \frac{c_9 (n+1)^2 \delta_0}{\chi \min\{\Lambda, (n+1)^\beta\}}.
\]
We take \(n = n_0\). Then on the one hand, we have
\[
(3.17) \quad \delta_{n_0} \leq \frac{c_9 (n_0+1)^2 \delta_0}{\chi \min\{\Lambda, (n_0+1)^\beta\}}.
\]
On the other hand,
\[
\delta_{n_0} = \frac{\delta_{n_0+1}}{H_{n_0}(m)^{m-1}} \geq \frac{\theta}{H_{n_0}(m)^{m-1}}.
\]
By Lemma 3.2 (ii), \(H_{n_0}(m) \leq m^{1/(m-1)} e^{\delta_{n_0}} \leq m^{1/(m-1)} e^\theta\), which is bounded by 3 by the choice of \(\theta\); hence
\[
E(m^{X_{n_0}}) \leq 3.
\]
We have \(\delta_{n_0} \geq \frac{\theta}{H_{n_0}(m)^{m-1}} \geq \frac{\theta}{m e^{(m-1)\beta}}\). Combining this with (3.17), we get
\[
\frac{(n_0+1)^2}{\chi \min\{\Lambda, (n_0+1)^\beta\}} \geq \frac{c_{10} \chi}{\delta_0},
\]
with \(c_{10} := \frac{\theta}{c_9 m e^{(m-1)\beta}}\). The proposition is proved (recalling from Remark 2.2 that \(\Lambda < \infty\) in case \(\beta = 2\)). \(\square\)

4. Open paths, XY-coupling, and the first crossing. The proof of the lower bound in Theorems 1.1, 1.2 and 1.3 requires deep probabilistic understanding of the model in the supercritical regime. It is made possible via a coupling argument, called XY-coupling, by means of a comparison between the supercritical system and an appropriate critical system. The coupling is quite elementary if we use the hierarchical representation of recursive systems.

The organization of this section is as follows. Section 4.1 introduces the hierarchical representation of recursive systems, regardless of whether they are supercritical or critical (or even subcritical, though we do not deal with subcritical systems in this paper). Section 4.2 gives the XY-coupling between the supercritical system and an appropriate critical system. In the brief but crucial Section 4.3, we introduce an important quantity for the critical system: the number of open paths. Section 4.4 is a bridge connecting the expected value in a supercritical system and the expected number of
open paths in a critical system. This bridge is crossed a first time in Section 4.5 to study a critical system by means of an appropriate supercritical system. [In Section 6, this bridge will be crossed a second time, in the opposite direction, to study a supercritical system by means of an appropriate critical system. Both crossings need much preparation: the first crossing relies on technical notions (hierarchical representation, open paths and the XY-coupling), whereas the preparation for the second crossing involves some technical estimates in Sections 5 and 6.]

4.1. The hierarchical representation. Let \((X_n, n \geq 0)\) be a system as defined in (1.1), i.e., such that each \(X_{n+1}\) has the law of \((X_{n,1} + \cdots + X_{n,m} - 1)^+\), where \(X_{n,k}, 1 \leq k \leq m\), are independent copies of \(X_n\). In order to introduce the XY-coupling between a supercritical system and a critical system, it turns out to be convenient to use a simple hierarchical representation of the system. As a matter of fact, it is in the form of the hierarchical representation that the system appeared in Collet et al. [5] and in Derrida and Retaux [8].

We define a family of random variables \((X(x), x \in T)\), indexed by a (reversed) \(m\)-ary tree \(T\), in the following way. For any vertex \(x\) in the genealogical tree \(T\), we use \(|x|\) to denote the generation of \(x\); so \(|x| = 0\) if \(x\) is in the initial generation. We assume that \(X(x)\), for \(x \in T\) with \(|x| = 0\) (i.e., in the initial generation of \(T\)), are i.i.d. having the distribution of \(X_0\). For any vertex \(x \in T\) with \(|x| \geq 1\), let \(x^{(1)}, \ldots, x^{(m)}\) denote the \(m\) parents of \(x\) in generation \(|x| - 1\), and set

\[
X(x) := (X(x^{(1)}) + \cdots + X(x^{(m)}) - 1)^+.
\]

See Figure 1.

![Figure 1](image_url)

**Figure 1.** An example of hierarchical representation (the first 4 generations) with \(m = 2\). Each vertex is represented by a circle; the number in the circle is the value of the system at the vertex.
As such, for any given \( n \geq 0 \), the sequence of random variables \( X(x) \), for \( x \in \mathbb{T} \) with \( |x| = n \), are i.i.d. having the distribution of \( X_n \).

The hierarchical representation is valid for any system satisfying the recursion (1.1), regardless of whether the system is supercritical, critical or subcritical.

4.2. The XY-coupling. Let \((X_n, n \geq 0)\) be a system as defined in (1.1). We are going to make a coupling, called XY-coupling, for \((X_n, n \geq 0)\) and an appropriate system \((Y_n, n \geq 0)\) such that the first system is always greater than or equal to the second in their hierarchical representation.

Assume \( \mathbb{P}(X_0 = k) \geq \mathbb{P}(Y_0 = k) \) for all integers \( k \geq 1 \), and \( \mathbb{P}(Y_0 = 0) > 0 \). We can couple the random variables \( X_0 \) and \( Y_0 \) in a same probability space such that \( X_0 \geq Y_0 \) a.s. and that \( \mathbb{P}(X_0 = Y_0 \mid Y_0 > 0) = 1 \).

For each \( n \geq 0 \), let \( Y_{n+1} \) be a random variable having the law of \((Y_{n,1} + \cdots + Y_{n,m} - 1)^+\), where \( Y_{n,1}, \ldots, Y_{n,m} \) are independent copies of \( Y_n \).

As in Section 4.1, let \( \mathbb{T} \) be the common genealogical m-ary tree associated with systems \((X_n, n \geq 0)\) and \((Y_n, n \geq 0)\). Let \((X(x), Y(x))\), for \( x \in \mathbb{T} \) with \( |x| = 0 \), be i.i.d. random variables having the distribution of \((X_0, Y_0)\), where \( X_0 \) and \( Y_0 \) are already coupled such that \( X_0 \geq Y_0 \) a.s. and that \( \mathbb{P}(X_0 = Y_0 \mid Y_0 > 0) = 1 \). For any vertex \( x \in \mathbb{T} \) with \( |x| \geq 1 \), let

\[
X(x) := (X(x^{(1)}) + \cdots + X(x^{(m)}) - 1)^+, \\
Y(x) := (Y(x^{(1)}) + \cdots + Y(x^{(m)}) - 1)^+, 
\]

where \( x^{(1)}, \ldots, x^{(m)} \) denote as before the parents of \( x \). It follows that for any given \( n \geq 0 \), the sequence of random variables \( X(x) \), indexed by \( x \in \mathbb{T} \) with \( |x| = n \), are i.i.d. having the distribution of \( X_n \), whereas the sequence of random variables \( Y(x) \), also indexed by \( x \in \mathbb{T} \) with \( |x| = n \), are i.i.d. having the distribution of \( Y_n \). Moreover, for all \( x \in \mathbb{T} \), \( X(x) \geq Y(x) \) a.s.

We observe that since \( X_0 \geq Y_0 \) a.s. and \( \mathbb{P}(X_0 = Y_0 \mid Y_0 > 0) = 1 \), we have

\[
\mathbb{E}(X_0 - Y_0 \mid Y_0 = 0) = \frac{\mathbb{E}[(X_0 - Y_0) \mathbb{1}_{\{Y_0 = 0\}}]}{\mathbb{P}(Y_0 = 0)} = \frac{\mathbb{E}(X_0 - Y_0)}{\mathbb{P}(Y_0 = 0)}.
\]

Here is a summary of the XY-coupling.

Theorem 4.1. (The XY-coupling). Assume \( \mathbb{P}(X_0 = k) \geq \mathbb{P}(Y_0 = k) \) for all integers \( k \geq 1 \), and \( \mathbb{P}(Y_0 = 0) > 0 \). We can couple two systems

\[\text{Let } Z \text{ be a } \mathbb{Z}_+ \text{-valued random variable, independent of } Y_0, \text{ such that } \mathbb{P}(Z = k) := \frac{\mathbb{P}(X_0 = k) - \mathbb{P}(Y_0 = k)}{\mathbb{P}(Y_0 = 0)} \text{ for } k \geq 1 \text{ and } \mathbb{P}(Z = 0) := \frac{\mathbb{P}(X_0 = 0)}{\mathbb{P}(Y_0 = 0)}. \text{ Then the pair } Y_0 + Z \mathbb{1}_{\{Y_0 = 0\}} \text{ and } Y_0 \text{ will do the job.}\]
(X(x), x ∈ T) and (Y(x), x ∈ T) with initial distributions X₀ and Y₀, respectively, such that X(x) ≥ Y(x) for all x ∈ T; in particular, we can couple two systems (Xₙ, n ≥ 0) and (Yₙ, n ≥ 0) such that Xₙ ≥ Yₙ a.s. for all n ≥ 0. Moreover, the coupling satisfies

\[
E(X₀ - Y₀ | Y₀ = 0) = \frac{E(X₀ - Y₀)}{P(Y₀ = 0)}.
\]

We are going to apply the XY-coupling several times. Each time, (Yₙ, n ≥ 0) is critical (so the condition P(Y₀ = 0) > 0 is automatically satisfied), and (Xₙ, n ≥ 0) supercritical.

4.3. Open paths. Let (Yₙ, n ≥ 0) denote a system satisfying (1.1). For any vertex x ∈ T, we call (xₖ, 0 ≤ k ≤ |x|) a path leading to x if each xₖ₊₁ is the (unique) child of xₖ, and |xₖ| = k. [Degenerate case: when |x| = 0, the path leading to x is reduced to the singleton x.] A path is said to be open if for any vertex x on the path with |x| ≥ 1, we have \(Y(x) = Y(x^{(1)}) + \cdots + Y(x^{(m)}) - 1\) (or equivalently, \(Y(x^{(1)}) + \cdots + Y(x^{(m)}) ≥ 1\)). [Degenerate case: when |x| = 0, the path is considered as open.]

For x ∈ T and integer i ≥ 0, let \(N^i(x)\) denote the number of open paths leading to x, and \(N^i(x)\) the number of open paths leading to x such that \(Y(x₀) = i\); note that an open path can start with a vertex x₀ in the initial generation with \(Y(x₀) = 0\) if it receives enough support from neighbouring vertices along the path. By definition, 0 ≤ \(N^i(x)\) ≤ \(N^i(x)\) ≤ \(m^{|x|}\) for all x ∈ T and integers i ≥ 0, and \(N^i(y) = 1\) and \(N^i(y) = 1_{\{Y(y) = i\}}\) for y ∈ T with |y| = 0.

For n ≥ 0, let \(eₙ\) denote the first lexicographic vertex in the n-th generation of T. See Figure 2. We write

\[
N^i_n := N^i(eₙ), \quad N^i_n := N^i(eₙ), \quad Y_n := Y(eₙ).
\]

[In the paper, only the law of each Yₙ individually is concerned; since Y(\(eₙ\)) has the same distribution as Yₙ, the abuse of notation \(Y_n := Y(eₙ)\), which has the advantage of making formulae and discussions more compact, should not be source of any confusion.] See Figure 2.

4.4. A bridge connecting two banks. Assume \(P(X₀ = k) ≥ P(Y₀ = k)\) for all integers k ≥ 1, and \(P(Y₀ = 0) > 0\). Let (Xₙ, n ≥ 0) and (Yₙ, n ≥ 0) be systems coupled via the XY-coupling (see Theorem 4.1). Thanks to the notion of the number of open paths \(N^0_n\) for (Yₙ) introduced in (4.2), we are
Fig 2. The same example as in Figure 1, with open paths marked in bold (and coloured in red). For this example, $N_4^# = 6$, $N_4^{(0)} = 2$, $N_4^{(1)} = 1$, $N_4^{(2)} = 0$, $N_4^{(3)} = 2$, $N_4^{(4)} = 1$, and $N_4^{(i)} = 0$ if $i \geq 5$.

now able to give a lower bound for $E(X_n)$ in terms of the number of open paths of $(Y_n)$.

**Theorem 4.2.** (The bridge inequality). Assume $P(X_0 = k) \geq P(Y_0 = k)$ for all integers $k \geq 1$, and $P(Y_0 = 0) > 0$. Let $(X_n, n \geq 0)$ and $(Y_n, n \geq 0)$ be systems coupled via the $XY$-coupling in Theorem 4.1. Let $\eta > 0$. If $E(X_0) < \infty$ and $E(X_0 - Y_0) \geq \eta P(Y_0 = 0)$, then for all $r \geq 0$, all integers $n \geq 0$, and $k \geq 0$ and $\ell \in [0, \frac{r\eta}{2}]$,

$$E(X_{n+k+\ell}) \geq \frac{m^{k+\ell} \eta}{2} E\left[N_n^{(0)} 1_{\{N_n^{(0)} \geq r\}} 1_{\{Y_n = k\}}\right].$$

**Proof.** Let $A_n$ denote the set of vertices $x \in T$ with $|x| = 0$ that are the beginning of an open path leading to $e_n$. [So in the notation of (4.2), the cardinality of $A_n$ is $N_n^\#$.] Let

$$A_n^{(0)} := \{x \in A_n : Y(x) = 0\}.$$ 

The cardinality of $A_n^{(0)}$ is $N_n^{(0)}$, defined in (4.2). Write

$$\xi_n := \sum_{x \in A_n^{(0)}} [X(x) - Y(x)].$$

[Recall that $X(x) \geq Y(x)$ a.s. and $P(X(x) = Y(x) | Y(x) > 0) = 1$ for all $x \in T$ with $|x| = 0$.] The crucial observation is that for all $n \geq 1$,

$$X(e_n) \geq Y(e_n) + \xi_n, \quad \text{a.s.}$$
So for any integers \( k \geq 0 \) and \( r \geq 0 \), and any \( \ell \geq 0 \),

\[
(X(\epsilon_n) - k - \ell)^+ \geq (Y(\epsilon_n) + \xi_n - k - \ell)^+ 1_{\{N_n^{(0)} \geq r\}} 1_{\{Y(\epsilon_n) = k\}}.
\]

= \((\xi_n - \ell)^+ 1_{\{N_n^{(0)} \geq r\}} 1_{\{Y(\epsilon_n) = k\}}\).

Let

\[\mathcal{Y} := \sigma(Y(x), x \in \mathbb{T}) = \sigma(Y(x), x \in \mathbb{T}, |x| = 0),\]

which is the sigma-field generated by the hierarchical system \((Y(x), x \in \mathbb{T})\).

Since \(\{N_n^{(0)} \geq r\} \in \mathcal{Y}\) and \(\{Y(\epsilon_n) = k\} \in \mathcal{Y}\), we obtain:

\[
E((X(\epsilon_n) - k - \ell)^+ | \mathcal{Y}) \geq 1_{\{N_n^{(0)} \geq r\}} 1_{\{Y(\epsilon_n) = k\}} E((\xi_n - \ell)^+ | \mathcal{Y}).
\]

Conditionally on \(x \in A_n^{(0)}\), the random variable \(X(x) - Y(x)\) has the same law as \(X_0 - Y_0\) conditionally on \(\{Y_0 = 0\}\); hence

\[
E(\xi_n | \mathcal{Y}) = N_n^{(0)} E(X_0 - Y_0 | Y_0 = 0) = \frac{N_n^{(0)} E(X_0 - Y_0)}{P(Y_0 = 0)},
\]

the last equality being a consequence of (4.1). By assumption, this yields \(E(\xi_n | \mathcal{Y}) \geq \eta N_n^{(0)}\). Consequently, for any \( \ell \geq 0 \),

\[
E((\xi_n - \ell)^+ | \mathcal{Y}) \geq E((\xi_n - \ell) | \mathcal{Y}) \geq \eta N_n^{(0)} - \ell.
\]

Combined with (4.4) and taking expectation, it follows that for any integers \( k \geq 0 \) and \( r \geq 0 \), and any \( \ell \geq 0 \),

\[
E[(X(\epsilon_n) - k - \ell)^+] \geq E\left[ (\eta N_n^{(0)} - \ell) 1_{\{N_n^{(0)} \geq r\}} 1_{\{Y(\epsilon_n) = k\}} \right].
\]

On the event \(\{N_n^{(0)} \geq r\}\), we have \(\eta N_n^{(0)} - \ell \geq \frac{\eta N_n^{(0)}}{2}\) if \( \ell \leq \frac{rn}{2} \). Hence, for any integers \( k \geq 0 \), \( r \geq 0 \) and \( 0 \leq \ell \leq \frac{rn}{2} \),

\[
E[(X(\epsilon_n) - k - \ell)^+] \geq E\left[ \frac{\eta}{2} N_n^{(0)} 1_{\{N_n^{(0)} \geq r\}} 1_{\{Y(\epsilon_n) = k\}} \right].
\]

If \( \ell \geq 0 \) is taken to be an integer, then obviously

\[
E(X(\epsilon_{n+k+\ell})) \geq m^{k+\ell} E[(X(\epsilon_n) - k - \ell)^+].
\]

Consequently, for any integers \( k \geq 0 \), \( r \geq 0 \) and \( \ell \in [0, \frac{rn}{2}] \),

\[
E(X(\epsilon_{n+k+\ell})) \geq \frac{m^{k+\ell} \eta}{2} E\left[ N_n^{(0)} 1_{\{N_n^{(0)} \geq r\}} 1_{\{Y(\epsilon_n) = k\}} \right].
\]

Recall that \(X(\epsilon_{n+k+\ell})\) has the distribution of \(X_{n+k+\ell}\), and that \(Y(\epsilon_n) = Y_n\) by notation (see (4.2)). This yields Theorem 4.2. \[\square\]
4.5. The first crossing. Theorem 4.2 yields the following estimate for the number of open paths in a critical system. This is our first crossing of the bridge, studying the critical system by means of a supercritical system. In Section 6, we are going to make a second crossing of the same bridge, but in the opposite direction, studying the supercritical system by means of a critical system.

As in (2.1), we keep using the following notation:

\[ \Lambda = \Lambda(Y_0) := \mathbb{E}(Y_0^\beta m^{Y_0}) \in (0, \infty], \]

associated with the critical system \((Y_n, n \geq 0)\), i.e., satisfying \((m - 1)\mathbb{E}(Y_0 m^{Y_0}) = \mathbb{E}(m^{Y_0}) < \infty\). Note that \(\mathbb{P}(Y_0 = 0) > 0\).\(^5\) Recall from (3.2) that for the critical system \((Y_n, n \geq 0)\), we have

\[ \Lambda \in \left[ -\frac{1}{m - 1}, \infty \right]. \]

Let \(\beta \in [0, 2]\) and \(\chi \in (0, 1]\). We assume the system \((Y_n, n \geq 0)\) is \(\beta\)-regular with coefficient \(\chi\) in the sense of (2.3). Recall from Remark 2.2 that \(\Lambda < \infty\) in case \(\beta = 2\). Write

\[ \lambda_n = \lambda_n(Y_0, \beta) := \min\{\Lambda, n^\beta\}. \]

**Corollary 4.3.** Let \(\beta \in [0, 2]\) and \(\chi \in (0, 1]\). Assume \((m - 1)\mathbb{E}(Y_0 m^{Y_0}) = \mathbb{E}(m^{Y_0}) < \infty\) and the system \((Y_n, n \geq 0)\) is \(\beta\)-regular with coefficient \(\chi\) in the sense of (2.3). Let \(N_n^{(0)}\) be as in (4.2). There exist constants \(c_{11} > 0\) and \(c_{12} > 0\), depending only on \(m\) and \(\beta\), such that for all \(n \geq 2\),

\[
\mathbb{E}\left[m^{Y_n}N_n^{(0)} 1_{\{N_n^{(0)} \geq \chi n^2 \log n^{-1} \lambda_n \leq n\}} 1\{Y_n \leq n\}\right] \leq \frac{c_{12}}{\chi n^2},
\]

where \(\lambda_n\) is defined in (4.5).

**Proof.** Fix \(n \geq 2\). Let \(c_2 \in (0, 1]\) and \(c_3 \in (0, 1]\) be the constants in Proposition 3.1. Let \(n_1 = 10n\) and \(\eta = \frac{c_2^{c_2} m \chi}{(m - 1)^2} \frac{\lambda_{10n}}{(10n)^2}\). Then \((m - 1)^2 \eta \in (0, c_3 \chi) \subset (0, c_3)\).

\(^5\) Recall that we always assume \(\mathbb{P}(Y_0 \geq 2) > 0\) to avoid triviality.
Let \((X_n^{(n)}, n \geq 0)\) be a system with \(P(X_0^{(n)} = k) = P(Y_0 = k)\) for \(k \geq 2\), and
\[
\begin{align*}
P(X_0^{(n)} = 0) &= \max\{P(Y_0 = 0) - \eta, 0\}, \\
P(X_0^{(n)} = 1) &= P(Y_0 = 1) + \min\{P(Y_0 = 0), \eta\}.
\end{align*}
\]
Then \((m-1)E(X_0^{(n)} m X_0^{(n)}) - E(m X_0^{(n)}) = (m-1)^2 \min\{P(Y_0 = 0), \eta\} \leq (m-1)^2 \eta\). Moreover, \(E((X_0^{(n)})^2 m X_0^{(n)}) = E(Y_0 m Y_0) + m \min\{P(Y_0 = 0), \eta\} = \Lambda + m \min\{P(Y_0 = 0), \eta\}\). Since \(\Lambda \geq \frac{1}{m-1}\), whereas \(m \min\{P(Y_0 = 0), \eta\} \leq m \eta \leq \frac{m}{(m-1)^2} \leq \frac{m}{m-1} \leq m \Lambda\), we get \(E((X_0^{(n)})^2 m X_0^{(n)}) \leq (m + 1)\Lambda\). By assumption, \((Y_n, n \geq 0)\) is \(\beta\)-regular with coefficient \(\chi\) in the sense of (2.3), so the supercritical system \((X_n^{(n)}, n \geq 0)\) is also \(\beta\)-regular with coefficient \(\chi / (m+1)\).

By definition of \(n_1\) and \(\eta\), we have \(n_1 \leq \min\{(\frac{c_2}{(m+1)^2} \Lambda)^{1/2}, (\frac{c_3}{(m+1)^2} \eta)^{1/2}\}\) for \(\beta \in [0, 2)\) and \(n_1 \leq (\frac{c_2}{(m+1)^2} \Lambda)^{1/2}\) for \(\beta = 2\). So we are entitled to apply Proposition 3.1 to get
\[
\max_{0 \leq i \leq n_1} E(X_{i}^{(n)}) \leq 3.
\]

On the other hand, \(P(X_0^{(n)} = k) \geq P(Y_0 = k)\) for all \(k \geq 1\), and \(P(Y_0 = 0) > 0\) as observed previously; since \(\frac{E(X_0^{(n)} - Y_0)}{P(Y_0 = 0)} = \min\{P(Y_0 = 0), \eta\} \geq \eta\), we are entitled to apply Theorem 4.2 to obtain, for all \(r \geq 0\), all integers \(k \geq 0\) and \(\ell \in [0, \frac{r \eta}{2}]\),
\[
E\left[N_n^{(0)} 1_{\{N_n^{(0)} \geq r\}} 1_{\{Y_n = k\}}\right] \leq \frac{2E(X_{n+k+\ell}^{(n)})}{m^{k+\ell} \eta}.
\]

Hence for all \(r \geq 0\), and all integers \(\ell \in [0, \frac{r \eta}{2}]\) with \(2n + \ell \leq n_1\), we have
\[
E\left[m Y_n N_n^{(0)} 1_{\{N_n^{(0)} \geq r\}} 1_{\{Y_n \leq n\}}\right] \leq \frac{2(n+1)}{m^r \eta} \max_{0 \leq i \leq n_1} E(X_{i}^{(n)}) \leq \frac{6(n + 1)}{m^r \eta}.
\]

We take \(\ell := \left\lfloor \frac{5 \log \eta}{\log m} \right\rfloor\), \(r := \frac{c_{11}}{\chi} n^2 \log n \frac{1}{\lambda_n}\) and \(c_{11} := \frac{10^3 (m-1)^2 (m+1)}{c_3 c_2 \log m}\). Then
\(2n + \ell \leq 10n\) (because \(\frac{5 \log \eta}{\log m} \leq \frac{5}{\log 2} < 8\)) and \(\ell \leq \frac{r \eta}{2}\). The corollary follows readily.

5. Preparation for the second crossing. Throughout the section, let \((Y_n, n \geq 0)\) denote a critical system, i.e., satisfying \((m-1)E(Y_0 m Y_0) = E(m Y_0) < \infty\).

The goal of this section is to establish a recursive formula (Proposition 5.1 below) for the weighted expected number of open paths.
The section is split into two parts. The first part gives a recursive formula for the number of open paths in the critical regime. The second part collects some known moment estimates for the system in the critical regime.

5.1. A recursive formula. Let \((Y_n, n \geq 0)\) be a system satisfying \((m - 1)E(Y_0^m Y_0) = E(m Y_0) < \infty\). The following proposition gives a recursive formula for the weighted expected number of open paths, where \(N_n\) can be \(N_n^{(i)}\) (for any integer \(i \geq 0\)), or \(N_n^\#\), as defined in (4.2). Note that by definition, \(E[m Y_0(1+Y_0)N_n^{(i)}] = m^i(1+i)P(Y_0 = i)\) and \(E[m Y_0(1+Y_0)N_n^\#] = E[m Y_0(1 + Y_0)] < \infty\).

In our paper, only the formula for \(N_n = N_n^{(0)}\) is of interest.

**Proposition 5.1.** (Recursive formula for number of open paths). Assume \((m - 1)E(Y_0^m Y_0) = E(m Y_0) < \infty\). Fix an integer \(i \geq 0\). Let \(N_n\) denote either \(N_n^\#\) or \(N_n^{(i)}\). Then

\[
E[m Y_n(1 + Y_n)N_n] = E[m Y_0(1 + Y_0)N_0] \prod_{k=0}^{n-1} [E(m Y_k)]^{m-1}, \quad n \geq 1.
\]

**Proof.** For any vertex \(x \in \mathbb{T}\) with \(|x| = n \geq 1\), let \(x^{(1)}, \ldots, x^{(m)}\) be as before its parents in generation \(n - 1\). By definition,

\[
Y(x) = [Y(x^{(1)}) + \cdots + Y(x^{(m)}) - 1]^+,
\]

\[
N(x) = [N(x^{(1)}) + \cdots + N(x^{(m)})] 1_{\{Y(x^{(1)}) + \cdots + Y(x^{(m)}) \geq 1\}},
\]

where \(N(y)\) can be either \(N^\#(y)\) or \(N^{(i)}(y)\). Write \(\Sigma_Y = \Sigma_Y(x) := Y(x^{(1)}) + \cdots + Y(x^{(m)})\) and \(\Sigma_N = \Sigma_N(x) := N(x^{(1)}) + \cdots + N(x^{(m)})\). Then \((Y(x), N(x)) = (\Sigma_Y - 1, \Sigma_N) 1_{\{\Sigma_Y \geq 1\}} + (0, 0) 1_{\{\Sigma_Y = 0\}}\). Consequently, for \(s > 0\) and \(t \geq 0\) (with \(0^0 := 1\)),

\[
E(s^Y t^N) = E(s^{\Sigma_Y - 1} t^{\Sigma_N} 1_{\{\Sigma_Y \geq 1\}}) + P(\Sigma_Y = 0)
\]

\[
= E(s^{\Sigma_Y - 1} t^{\Sigma_N}) - s^{-1} E(t^{\Sigma_N} 1_{\{\Sigma_Y = 0\}}) + P(\Sigma_Y = 0).
\]

For further use, we observe that the same argument yields \(P(N(x) \geq 1, Y(x) = 0) = P(\Sigma_N \geq 1, \Sigma_Y = 1)\), which is bounded by \(P(\Sigma_Y \geq 1)\). Hence

\[
P(N(x) \geq 1, Y(x) = 0) \leq m P(Y(x^{(1)}) \geq 1).
\]

Define

\[
G_n(s, t) := E(s^Y t^N_n), \quad s \geq 0, \quad t \geq 0,
\]
which is the joint moment generating function for the pair \((Y_n, N_n)\). Since \((Y(x(i)), N(x(i)))\), for \(1 \leq i \leq m\), are i.i.d. having the distribution of \((Y_{n-1}, N_{n-1})\), we have \(E(s^{\Sigma Y}t^{\Sigma N}) = s^{-1}G_{n-1}(s, t)^m\), \(E(t^{\Sigma N}1_{\{\Sigma Y=0\}}) = [E(t^{N_{n-1}1_{\{Y_{n-1}=0\}}})]_m = G_{n-1}(0, t)^m\), and \(P(\Sigma Y = 0) = [P(Y_{n-1} = 0)]_m = G_{n-1}(0, 1)^m\). So (5.1) reads: for \(n \geq 1\), \(s > 0\) and \(t \geq 0\),

\[
G_n(s, t) = \frac{G_{n-1}(s, t)^m}{s} - \frac{G_{n-1}(0, t)^m}{s} + G_{n-1}(0, 1)^m.
\]

For further use, we also rewrite (5.2) as follows:

\[
P(N_n \geq 1, Y_n = 0) \leq mP(Y_{n-1} \geq 1).
\]

Let

\[
I_n(s) := E(s^{Y_n}N_n),
\]

which is the partial derivative of \(G_n(s, t)\) with respect to \(t\) at 1. Then

\[
I_{n+1}(s) = \frac{m}{s}I_n(s)G_n(s)^{m-1} - \frac{m}{s}I_n(0)G_n(0)^{m-1},
\]

where

\[
G_n(s) := E(s^{Y_n}).
\]

[So \(G_n(s) = G_n(s, 1)\).] Multiplying by \(s\) on both sides of (5.6), and differentiating with respect to \(s\), we get

\[
I_{n+1}(s) + sI'_{n+1}(s) = m(m - 1)G_n(s)^{m-2}G'_n(s)I_n(s) + mG_n(s)^{m-1}I'_n(s).
\]

In the critical regime, we have \(m(m - 1)G'_n(m) = G_n(m)\) (see (5.10)), which yields, for \(n \geq 0\),

\[
I_{n+1}(m) + mI'_{n+1}(m) = G_n(m)^{m-1}(I_n(m) + mI'_n(m)).
\]

Iterating the identity, and noting that \(E[m^{Y_n}(1 + Y_n)N_n] = I_n(m) + mI'_n(m)\), we get

\[
E[m^{Y_n}(1 + Y_n)N_n] = E[m^{Y_0}(1 + Y_0)N_0] \prod_{k=0}^{n-1} G_k(m)^{m-1}.
\]

Proposition 5.1 is proved. \(\square\)
5.2. Moment estimates. The assumption \((m - 1)\mathbb{E}(Y_0 m^{Y_0}) = \mathbb{E}(m^{Y_0}) < \infty\) implies (Collet et al. [6]) that for all \(n \geq 0\),
\[
(m - 1)\mathbb{E}(Y_n m^{Y_n}) = \mathbb{E}(m^{Y_n}) < \infty.
\]
By (3.11), with \(c_{13} := (1 + \frac{1}{m-1})m^{1/(m-1)}\),
\[
\mathbb{E}[(1 + Y_n)m^{Y_n}] \leq c_{13}, \quad n \geq 0.
\]
We keep using the notation introduced previously:
\[\Lambda = \Lambda(Y_0) := \mathbb{E}(Y_0^3 m^{Y_0}) \in \left[\frac{1}{m-1}, \infty\right],\]
\[\lambda_n = \lambda_n(Y_0) := \min\{\Lambda, n^\beta\}.\]
The next is a collection of known useful properties concerning the moments of \(Y_n\).

**Fact 5.2.** ([4]) Let \(\beta \in [0, 2] \) and \(\chi \in (0, 1]\). Assume \((m - 1)\mathbb{E}(Y_0 m^{Y_0}) = \mathbb{E}(m^{Y_0}) < \infty\). Assume the system \((Y_n, n \geq 0)\) is \(\beta\)-regular with coefficient \(\chi\) in the sense of (2.3). Then there exist constants \(c_{14} > 0, c_{15} > 0, c_{16} > 0\) and \(c_{17} > 0\), depending only on \(m\) and \(\beta\), such that for all \(n \geq 1\),
\[
\frac{c_{14}}{\Lambda} n^2 \leq \prod_{k=0}^{n-1} \mathbb{E}(m^{Y_k}) \leq \frac{c_{15}}{\chi} \frac{n^2}{\lambda_n},
\]
\[
\mathbb{E}(Y_n^2 m^{Y_n}) \leq c_{16} \left(\frac{\Lambda}{\chi} \frac{n^2}{\lambda_n}\right)^{1/2},
\]
\[
\mathbb{E}(Y_n^3 m^{Y_n}) \leq c_{17} \left(\frac{\Lambda}{\chi} \frac{n^2}{\lambda_n}\right)^{1/2}.
\]
The first inequality in (5.12) follows from Proposition 2 in [4], whereas the second inequality follows from the proof of Proposition 1 in [4] but by replacing the last displayed inequality in the proof (saying that \(\Delta_0(m - \frac{m}{n+2}) \geq \frac{P(X_0=0)}{2(n+2)^2}\) in the notation of [4]) by the assumption of the \(\beta\)-regularity in the sense of (2.3). Inequality (5.14) follows from (19) in [4] and the second inequality in (5.12). Finally, (5.13) is a consequence of (5.11), (5.14) and the Cauchy–Schwarz inequality. All the inequalities are valid even in case \(\Lambda = \infty\).
Remark 5.3. Consider a system \((Y_n, n \geq 0)\) satisfying \(((m-1)E(Y_0 m^{Y_0}) = E(m^{Y_0}) < \infty\). Assume \((Y_n, n \geq 0)\) is \(\beta\)-regular (for some \(\beta \in [0, 2]\)) with coefficient \(\chi \in (0, 1]\) in the sense of (2.3).

Combining Proposition 5.1 with (5.12) yields that there exist constants \(c_{18} > 0\) and \(c_{19} > 0\), depending only on \(m\), such that

\[
(5.15) \quad c_{18} n^2 \frac{P(Y_0 = 0)}{\Lambda} \leq E[m^{Y_0}(1 + Y_n)N_n^{(0)}] \leq \frac{c_{19} n^2 P(Y_0 = 0)}{\lambda_n},
\]

for all \(n \geq 1\). This is valid without assuming \(\Lambda\) to be finite.

About the factor \(P(Y_0 = 0)\) on both sides of (5.15): the assumption \(((m-1)E(Y_0 m^{Y_0}) = E(m^{Y_0}) < \infty\) implies \(P(Y_0 = 0) > 0\); in case \(m \geq 3\), we have, moreover, \(P(Y_0 = 0) \geq \frac{m-2}{m-1}\) (Lemma 3 in [4]). \(\square\)

6. The second crossing and a general lower bound. In Section 4.5, we crossed the bridge (Theorem 4.2) for the first time to obtain an upper bound for the number of open paths in a critical system. Now, when the recursive formula in Proposition 5.1 is established, we are going to cross again the bridge, but in the opposite direction, to obtain a lower bound for the expected value in a supercritical system. The second crossing will lead to a general lower bound for the free energy of certain systems that are not necessarily integer-valued.

Let \(\beta \in [0, 2]\) and \(\chi \in (0, 1]\). Let \((Y_n, n \geq 0)\) be a system satisfying \(((m-1)E(Y_0 m^{Y_0}) = E(m^{Y_0}) < \infty\) and is \(\beta\)-regular with coefficient \(\chi\) in the sense of (2.3). Let as before,

\[
\Lambda = \Lambda(Y_0) := E(Y_0^3 m^{Y_0}) \in \left[\frac{1}{m-1}, \infty\right],
\]

\[
\lambda_n = \lambda_n(Y_0) := \min\{\Lambda, n^\beta\}.
\]

By Remark 2.2, we have \(\Lambda < \infty\) in case \(\beta = 2\).

We start with a couple of lemmas.

Lemma 6.1. Let \(\beta \in [0, 2]\) and \(\chi \in (0, 1]\). Assume \(((m-1)E(Y_0 m^{Y_0}) = E(m^{Y_0}) < \infty\) and the system \((Y_n, n \geq 0)\) is \(\beta\)-regular with coefficient \(\chi\) in the sense of (2.3). Let \(N_n^{(0)}\) be as in (4.2). There exists a constant \(c_{20} > 0\), depending only on \(m\) and \(\beta\), such that for all integer \(n \geq 2\),

\[
(6.1) \quad E(m^{Y_n} N_n^{(0)}) \leq \frac{c_{20} n^2}{\chi} \frac{1}{\lambda_n} \left(\frac{1}{n} + (\log n) \left[E(m^{Y_n} 1_{\{Y_n \geq 1\}}) + E(m^{Y_{n-1}} 1_{\{Y_{n-1} \geq 1\}})\right]\right).
\]
Proof. Let $c_{11} > 0$ be the constant in Corollary 4.3. Write
\[ \mathbb{E}(m^{Y_n}N_n^{(0)}) \leq J_1(n) + J_2(n) + J_3(n), \]
where
\begin{align*}
J_1(n) &:= \mathbb{E}[m^{Y_n}N_n^{(0)} 1_{\{N_n^{(0)} > \frac{c_{11}}{\chi} n^2 \log n \}} 1_{\{Y_n < n\}}], \\
J_2(n) &:= \mathbb{E}[m^{Y_n}N_n^{(0)} 1_{\{Y_n \geq n\}}], \\
J_3(n) &:= \mathbb{E}[m^{Y_n}N_n^{(0)} 1_{\{N_n^{(0)} \leq \frac{c_{11}}{\chi} n^2 \log n \}} \leq \frac{c_{11}}{\chi} n^2 \log n \mathbb{E}[m^{Y_n} 1_{\{N_n^{(0)} \geq 1\}}].
\end{align*}

By Corollary 4.3, $J_1(n) \leq \frac{c_{11}}{\chi} n \leq \frac{c_{11}}{\chi} n^{1-\beta}$.

For $J_2(n)$, things are simple, because $m^{Y_n}N_n^{(0)} 1_{\{Y_n \geq n\}} \leq \frac{1}{n} Y_n m^{Y_n}N_n^{(0)}$, so by (5.15), for $n \geq 1$,
\[ J_2(n) \leq \frac{c_{19}}{\chi} \frac{n \mathbb{P}(Y_0 = 0)}{\lambda_n} \leq \frac{c_{19}}{\chi} \frac{n}{\lambda_n}. \]

For $J_3(n)$, we argue that $\mathbb{E}[m^{Y_n} 1_{\{N_n^{(0)} \geq 1\}}] \leq \mathbb{P}(N_n^{(0)} \geq 1, Y_n = 0) + \mathbb{E}[m^{Y_n} 1_{\{Y_n \geq 1\}}].$ Recall from (5.4) that $\mathbb{P}(N_n^{(0)} \geq 1, Y_n = 0) \leq m \mathbb{P}(Y_{n-1} \geq 1)$, which is bounded by $\mathbb{E}[m^{Y_{n-1}} 1_{\{Y_{n-1} \geq 1\}}].$ Hence
\[ J_3(n) \leq \frac{c_{11}}{\chi} \frac{n^2 \log n}{\lambda_n}[\mathbb{E}(m^{Y_n} 1_{\{Y_n \geq 1\}}) + \mathbb{E}(m^{Y_{n-1}} 1_{\{Y_{n-1} \geq 1\}})]. \]

Assembling these pieces yields the desired result. \hfill \square

For the next lemma, let us write $I_n(s) := \mathbb{E}(s^{Y_n}N_n^{(0)})$ as in (5.5).

**Lemma 6.2.** Let $\beta \in [0, 2]$ and $\chi \in (0, 1]$. Assume $(m - 1)\mathbb{E}(Y_0 m^{Y_0}) = \mathbb{E}(m^{Y_0}) < \infty$ and the system $(Y_n, n \geq 0)$ is $\beta$-regular with coefficient $\chi$ in the sense of (2.3). There exists a constant $c_{21} > 0$, depending only on $m$ and $\beta$, such that for all integer $n \geq 2$,
\[ \sum_{\ell=0}^{n} I_{\ell}(m) \leq \frac{c_{21}}{\chi^2} \frac{n^2 (\log n)^2}{\lambda_n}. \]
Proof. By Lemma 6.1,

\[
\sum_{\ell=0}^{n} I_{\ell}(m) \leq \mathbf{E}(m^{Y_0} N_0^{(0)}) + \mathbf{E}(m^{Y_1} N_1^{(0)}) + \sum_{\ell=2}^{n} \frac{c_20}{\chi} \left( \frac{\ell}{\lambda_{\ell}} + \frac{\ell^2 \log \ell}{\lambda_{\ell}} \left[ \mathbf{E}(m^{Y_{\ell}} 1_{\{Y_{\ell} \geq 1\}}) + \mathbf{E}(m^{Y_{\ell-1}} 1_{\{Y_{\ell-1} \geq 1\}}) \right] \right).
\]

On the right-hand side, \( N_0^{(0)} \leq 1 \) and \( N_1^{(0)} < m \), so \( \mathbf{E}(m^{Y_0} N_0^{(0)}) + \mathbf{E}(m^{Y_1} N_1^{(0)}) \leq \mathbf{E}(m^{Y_0}) + m \mathbf{E}(m^{Y_1}) \leq (m+1)c_{13} \) (by (5.11)). On the other hand, \( \frac{c_2}{\lambda_{\ell}} \leq \frac{\chi^2}{n} \).

Hence

\[
\sum_{\ell=0}^{n} I_{\ell}(m) \leq (m+1)c_{13} + \frac{c_20}{\chi} \frac{n^2}{\lambda_n} \sum_{\ell=2}^{n} \mathbf{E}(m^{Y_{\ell}} 1_{\{Y_{\ell} \geq 1\}}) + \mathbf{E}(m^{Y_{\ell-1}} 1_{\{Y_{\ell-1} \geq 1\}}) + 2c_20 \frac{n^2 \log n}{\chi} \sum_{\ell=0}^{n} \mathbf{E}(m^{Y_{\ell}} 1_{\{Y_{\ell} \geq 1\}}).
\]

(6.2)

On the other hand, noting that \( G_{\ell}(m) \leq m^{1/(m-1)} \) (see (3.11)) and there exists \( c_{22} = c_{22}(m) < \infty \) such that \( \log x \geq c_{22}(x-1) \) for \( x \in [1, m^{1/(m-1)}] \), we get

\[
\mathbf{E}(m^{Y_{\ell}} 1_{\{Y_{\ell} \geq 1\}}) \leq 2\mathbf{E}[(m^{Y_{\ell}} - 1) 1_{\{Y_{\ell} \geq 1\}}] = 2[\mathbf{E}(m^{Y_{\ell}}) - 1] \leq \frac{2}{c_{22}} \log[G_{\ell}(m)].
\]

Hence

\[
\sum_{\ell=0}^{n} \mathbf{E}(m^{Y_{\ell}} 1_{\{Y_{\ell} \geq 1\}}) \leq \frac{2}{c_{22}} \sum_{\ell=0}^{n} \log[G_{\ell}(m)] = \frac{2}{(m-1)c_{22}} \log \left( \prod_{\ell=0}^{n} G_{\ell}(m)^{m-1} \right) \leq \frac{2}{(m-1)c_{22}} \log \left( \frac{c_{15} (n+1)^2}{\chi \lambda_{n+1}} \right),
\]

the last inequality being a consequence of (5.12). Note that \( \log(\frac{c_{15} (n+1)^2}{\chi \lambda_{n+1}}) \leq c_{23} \frac{\log n}{\chi} \) for some constant \( c_{23} > 0 \) depending only on \( m, \beta \) and \( c_{15} \) (thus only on \( m \) and \( \beta \) once \( c_{15} \) is chosen in (5.12)). Substituting this into (6.2), we get the desired result. \( \square \)
Proposition 6.3. Let $\beta \in [0, 2]$ and $\chi \in (0, 1]$. Assume $(m - 1)E(Y_0 m Y_0) = E(m Y_0) < \infty$ and the system $(Y_n, n \geq 0)$ is $\beta$-regular with coefficient $\chi$ in the sense of (2.3). Let $N_n^{(0)}$ be as in (4.2). Then there exists a constant $c_{24} > 0$, depending only on $m$ and $\beta$, such that for all $n \geq 2$,

$$
E[(1 + Y_n)^3 m Y_n N_n^{(0)}] \leq c_{24} \frac{\Lambda}{\chi^3} \frac{n^4 (\log n)^2}{\Lambda_n^2}.
$$

Proof. Let $G_n(s) := E(s^Y_n)$ as in (5.7). Let $I_n(s) := E(s^Y_n N_n^{(0)})$ as in (5.5). [In particular, $I_0(s) = P(Y_0 = 0).$] Recall the iteration formula (5.8):

$$
I_{n+1}(s) + s I_n''(s) = m(m - 1)G_n(s) m^{-2} G_n''(s) I_n(s) + m G_n(s) m^{-1} I_n'(s).
$$

Differentiating twice on both sides with respect to $s$ and taking $s = m$, and using the identity $(m - 1)m G_n'(m) = G_n(m)$ in the critical regime (see (5.10)), we deduce that for some unimportant positive constants $c_{25} > 0$, $\ldots$, $c_{29} > 0$ and all $n \geq 1$, (6.3)

$$
3I_{n+1}''(m) + m I_n''(m) = G_n(m) m^{-1} \Upsilon_n + G_n(m) m^{-1} [3I_n''(m) + m I_n''(m)],
$$

where

$$
\Upsilon_n := c_{25} I_n(m) + c_{26} \frac{G_n''(m)}{G_n(m)} I_n(m) + c_{27} \frac{G_n'''(m)}{G_n(m)} I_n(m)
$$

$$
+ c_{28} I_n'(m) + c_{29} \frac{G_n'(m)}{G_n(m)} I_n'(m)
$$

$$
\leq c_{25} I_n(m) + c_{26} \frac{G_n'(m)}{G_n(m)} I_n(m) + c_{27} \frac{G_n''(m)}{G_n(m)} I_n(m)
$$

$$
+ c_{28} I_n'(m) + c_{29} G_n''(m) I_n'(m).
$$

At this stage, it is convenient to introduce

$$
\mathcal{D}_n(m) := m(m - 1)G_n'''(m) + (4m - 5)G_n''(m) + \frac{2(m - 2)}{m^2(m - 1)} G_n(m).
$$

Obviously, $\mathcal{D}_n(m) \geq c_{30} (1 + G_n''(m) + G_n'''(m))$ for some constant $c_{30} > 0$ depending only on $m$, so that with $c_{31} := c_{30} \max\{c_{25}, c_{26}, c_{27}\}$ and $c_{32} := c_{28} + c_{29},$

$$
\Upsilon_n \leq c_{31} I_n(m) \mathcal{D}_n(m) + c_{32} \max\{1, G_n''(m)\} I_n'(m)
$$

$$
\leq c_{31} I_n(m) \mathcal{D}_n(m) + c_{32} \max\{1, G_n''(m)\} [I_n(m) + m I_n'(m)].
$$
The interest of $D_n(m)$ lies in its recursive relation $D_{n+1}(m) = D_n(m) G_n(m)^{m-1}$ (see [4]), which leads to

$$D_n(m) = D_0(m) \prod_{i=0}^{n-1} G_i(m)^{m-1} \leq c_{33} \Lambda \prod_{i=0}^{n-1} G_i(m)^{m-1},$$

where $c_{33} > 0$ is a constant depending only on $m$. On the other hand, by (5.9),

$$I_n(m) + mI'_n(m) = (I_0(m) + mI'_0(m)) \prod_{i=0}^{n-1} G_i(m)^{m-1}.$$

Since $I_0(s) = P(Y_0 = 0)$ for all $s$, we have $I_0(m) + mI'_0(m) = P(Y_0 = 0) \leq 1$, which implies $I_n(m) + mI'_n(m) \leq \prod_{i=0}^{n-1} G_i(m)^{m-1}$. Consequently,

$$\Upsilon_n \leq (c_{34} \Lambda I_n(m) + c_{34} \max\{1, G''_n(m)\}) \prod_{i=0}^{n-1} G_i(m)^{m-1},$$

with $c_{34} := \max\{c_{31} c_{33}, c_{32}\}$. So we can rewrite (6.3) as

\begin{align*}
3I''_{n+1}(m) + m I'''_{n+1}(m) \\
\leq c_{34}(\Lambda I_n(m) + \max\{1, G''_n(m)\}) \prod_{i=0}^{n} G_i(m)^{m-1} \\
+ G_n(m)^{m-1}(3I''_n(m) + m I'''_n(m)).
\end{align*}

Iterating the inequality, and noting that $I''_0(m) = I'''_0(m) = 0$ (because $I_0(s) = P(Y_0 = 0)$ for all $s$), we obtain:

$$3I''_n(m) + m I'''_n(m) \leq c_{34} \left( \prod_{k=0}^{n-1} G_k(m)^{m-1} \right) \sum_{\ell=0}^{n-1} (\Lambda I_\ell(m) + \max\{1, G''_\ell(m)\}).$$

By (5.12), $\prod_{k=0}^{n-1} G_k(m)^{m-1} \leq \frac{c_{21}}{\chi} \frac{n^2}{\lambda_n^2}$, whereas by Lemma 6.2, $\sum_{\ell=0}^{n-1} I_\ell(m) \leq \frac{c_{21} n^2 (\log n)^2}{\lambda_n^2}$; hence

$$3I''_n(m) + m I'''_n(m) \leq \frac{c_{34} c_{15}}{\chi} \frac{n^2}{\lambda_n} \left( \Lambda \frac{c_{21} n^2 (\log n)^2}{\lambda_n^2} \right) + \sum_{\ell=0}^{n-1} \max\{1, G''_\ell(m)\}.$$

By (5.13), $G''_\ell(m) \leq \frac{c_{41}}{\chi^{1/2}} \left( \frac{\ell^2}{\chi} \Lambda \right)^{1/2} = \frac{c_{41}}{\chi^{1/2}} \max\{\ell, \Lambda^{1/2} \ell^{1-(\beta/2)}\} \leq \frac{c_{41}}{\chi^{1/2}} (n + \ldots$
$\Lambda^{1/2} n^{1-(\beta/2)}$; thus
\[
\sum_{\ell=0}^{n-1} \max\{1, G''_\ell(m)\} \leq \sum_{\ell=0}^{n-1} \left( 1 + \frac{c_{16}}{\chi^{1/2}} \left( n + \Lambda^{1/2} n^{1-(\beta/2)} \right) \right)
\]
\[
= n + \frac{c_{16}}{\chi^{1/2}} \left( n^2 + \Lambda^{1/2} n^{2-(\beta/2)} \right)
\]
\[
\leq \frac{1 + c_{16}}{\chi^{1/2}} \left( n^2 + \Lambda^{1/2} n^{2-(\beta/2)} \right).
\]

Since $n^2 + \Lambda^{1/2} n^{2-(\beta/2)} \leq 2 \max\{\Lambda^{1/2} n^{2-(\beta/2)}, n^2\} \leq 2 \max\{\Lambda n^{2-\beta}, n^2\} = 2 n^2 \Lambda$, this yields
\[
\sum_{\ell=0}^{n-1} \max\{1, G''_\ell(m)\} \leq \frac{2(1 + c_{16})}{\chi^{1/2}} \frac{n^2}{\Lambda}.
\]

Consequently,
\[
3I''(m) + mI''(m) \leq \frac{c_{34} c_{15}}{\chi} \frac{n^2}{\lambda_n} \left( \Lambda \frac{c_{21} n^2 (\log n)^2}{\lambda_n^2} + \frac{2(1 + c_{16}) n^2}{\chi^{1/2} \lambda_n^2} \right),
\]
proving Proposition 6.3 (recalling that $\chi \leq 1$).

**Proposition 6.4.** Let $\beta \in [0, 2]$ and $\chi \in (0, 1]$. Assume $\Lambda := \mathbb{E}(Y_0^3 m Y_0) < \infty$. Assume $(m-1)\mathbb{E}(Y_0^2 m Y_0) = \mathbb{E}(m Y_0)$ and the system $(Y_n, n \geq 0)$ is $\beta$-regular with coefficient $\chi$ in the sense of (2.3). Let $N_n^{(0)}$ be as in (4.2). There exist positive constants $c_{35}, c_{36}$ and $c_{37}$ which depend only on $(m, \beta)$ such that, for all $n \geq 2$,
\[
(6.4) \quad \mathbb{E}\left[ (1 + Y_n) m Y_n \mathbb{I}_{N_n^{(0)}, (N_n^{(0)})^2 > c_{35} n^2 \mathbb{P}(Y_0 = 0) \mathbb{I}_{Y_n < \frac{c_{36}}{\chi^{1/2} \lambda_n} \mathbb{I}_{\mathbb{P}(Y_0 = 0) \mathbb{I}_{\mathbb{P}(Y_0 = 0)^{1/2}}}}) \right]
\]
\[
\geq c_{37} \frac{n^2 \mathbb{P}(Y_0 = 0)}{\Lambda},
\]
Consequently, there exist a constant $c_{38} > 0$ depending only on $(m, \beta)$, and an integer $k$ with $0 \leq k < c_{38} \frac{c_{36}}{\chi^{1/2} \lambda_n} \mathbb{P}(Y_0 = 0)^{1/2}$, such that
\[
(6.5) \quad \mathbb{E}\left[ N_n^{(0)} \mathbb{I}_{N_n^{(0)}, (N_n^{(0)})^2 > c_{35} n^2 \mathbb{P}(Y_0 = 0) \mathbb{I}_{Y_n = k}} \right] \geq c_{38} \frac{\chi^2}{m^k \Lambda^3} \frac{\lambda_n^2 \mathbb{P}(Y_0 = 0)^2}{(\log n)^2}.
\]
Proof. Only (6.5) needs to be proved. Let $c_{39} > 0$ and let $K \geq 1$ be an integer; their values will be chosen later. By Proposition 6.3,

$$
E[(1 + Y_n)m^{Y_n} N_n(0) 1_{\{Y_n \geq K\}}] \leq \frac{E[(1 + Y_n)^3 m^{Y_n} N_n(0)]}{K^2} \leq \frac{c_{24} n^4 A (\log n)^2}{\lambda_n^2}.
$$

On the other hand, $N_n(0) 1_{\{N_n(0) \leq c_{39} n^2 P(Y_0=0) / \lambda\}} \leq c_{39} n^2 P(Y_0=0) / \lambda$, so by (5.11),

$$
E[(1 + Y_n)m^{Y_n} N_n(0) 1_{\{N_n(0) \leq c_{39} n^2 P(Y_0=0) / \lambda\}}] \leq c_{39} n^2 P(Y_0=0) / \lambda E[(1 + Y_n)m^{Y_n}]
$$

Recall from (5.15) that $E[(1 + Y_n)m^{Y_n} N_n(0)] \geq c_{18} n^2 P(Y_0=0) / \lambda$. Hence

$$
E[(1 + Y_n)m^{Y_n} N_n(0) 1_{\{N_n(0) > c_{39} n^2 P(Y_0=0) / \lambda\}} 1_{\{Y_n < K\}}] \geq E[(1 + Y_n)m^{Y_n} N_n(0)] - E[(1 + Y_n)m^{Y_n} N_n(0) 1_{\{Y_n \geq K\}}] - E[(1 + Y_n)m^{Y_n} N_n(0) 1_{\{N_n(0) \leq c_{39} n^2 P(Y_0=0) / \lambda\}}]
$$

Choosing $c_{39} := \frac{c_{18}}{3c_{13}}$ and $K := \lfloor \frac{3c_{24}}{c_{18}} \rfloor^{1/2} \chi^{-3/2} \frac{n \lambda \log n}{\lambda_n [P(Y_0=0)]^{1/2}} + 1$ yields (6.5). Proposition 6.4 is proved.

We are now ready for the second crossing of the bridge which is Theorem 4.2. It implies a general lower bound for the free energy of certain supercritical systems $(X_n, n \geq 0)$. We are also going to apply this general result to prove the lower bounds in Theorem 1.1 (Section 7) and in Theorems 1.2 and 1.3 (Section 8).

Assume $(m - 1)E(Y_0 m^{Y_0}) = E(m^{Y_0}) < \infty$. In particular, $P(Y_0 = 0) > 0$. Assume $P(X_0 = k) \geq P(Y_0 = k)$ for all integers $k \geq 1$. We can couple two systems $(X_n, n \geq 0)$ and $(Y_n, n \geq 0)$ via the $XY$-coupling in Theorem 4.1.

Theorem 6.5. Let $\beta \in [0, 2]$ and $\chi \in (0, 1]$. Assume $\Lambda = \Lambda(Y_0) := E(Y_0^3 m^{Y_0}) < \infty$. Assume $(m - 1)E(Y_0 m^{Y_0}) = E(m^{Y_0})$ and the system $(Y_n, n \geq 0)$ is $\beta$-regular with coefficient $\chi$ in the sense of (2.3). Assume $P(X_0 = k) \geq P(Y_0 = k)$ for all integers $k \geq 1$. Let $(X_n, n \geq 0)$ and $(Y_n, n \geq 0)$ be systems coupled via the $XY$-coupling in Theorem 4.1. There exists a
constant $c_{40} > 0$, depending only on $m$ and $\beta$, such that if $\mathbf{E}(X_0 - Y_0) \geq \eta$ for some $\eta \in (0, \frac{1}{d(m-1)})$,\(^6\) then

\[(6.6) \quad \max_{0 \leq j \leq n_0} \mathbf{E}(X_j) \geq 2,\]

where $n_0 := \lfloor \chi^3 \mathbf{P}(Y_0 = 0)^{1/2} \max \{ \Lambda (R \log R)^{(1-\beta)/2}, (R \log R)^{1/2} \log R \} \rfloor / 2$ with $R := \frac{\Lambda}{n \chi \mathbf{P}(Y_0 = 0)}$; in particular, the free energy $F^X$ of the system $(X_n, n \geq 0)$ satisfies

\[(6.7) \quad F^X \geq \exp \left[ -\frac{c_{41}}{\chi^3 \mathbf{P}(Y_0 = 0)^{1/2}} \max \{ \Lambda (R \log R)^{(1-\beta)/2}, (R \log R)^{1/2} \log R \} \right],\]

where $c_{41} > 0$ is a constant whose value depends only on $m$ and $\beta$.

Proof. Only (6.6) needs proving, because (6.7) follows immediately from (6.6) by means of (1.5).

Let $((X(x), Y(x)), x \in \mathbb{T})$ be the pair of systems in the XY-coupling in Theorem 4.1 satisfying $\mathbf{E}(X_0 - Y_0) \geq \eta$. Recall from Theorem 4.2 that for all $r \geq 0$, all integers $n \geq 3$, $k \geq 0$ and $\ell \in [0, \frac{n \Lambda \log n}{\chi \mathbf{P}(Y_0 = 0)^{1/2}}],\(^7\)

\[
\mathbf{E}(X_{n+k+\ell}) \geq \frac{m_{k+\ell} \eta}{2 \mathbf{P}(Y_0 = 0)} \mathbf{E} \left[ N_n \left\{ \chi^3 \mathbf{P}(Y_0 = 0)^{1/2} \right\} \mathbf{1}\{Y_n = k\} \right].
\]

By (6.5) in Proposition 6.4, there exists an integer $k$ with $0 \leq k < \frac{c_{36}}{\chi \mathbf{P}(Y_0 = 0)^{1/2}}$, such that

\[
\mathbf{E} \left[ N_n \left\{ \chi^3 \mathbf{P}(Y_0 = 0)^{1/2} \right\} \mathbf{1}\{Y_n = k\} \right] \geq c_{38} \frac{\chi^3}{m^k \Lambda^3} \frac{\Lambda^2 \mathbf{P}(Y_0 = 0)^2}{(\log n)^2},
\]

which implies that for some $0 \leq k < \frac{c_{36}}{\chi \mathbf{P}(Y_0 = 0)^{1/2}}$, all $\ell \in [0, \frac{c_{38} n^2 \Lambda}{2 \eta}],\(^7\)

\[
\mathbf{E}(X_{n+k+\ell}) \geq \frac{m_{k+\ell} \eta}{2 \mathbf{P}(Y_0 = 0)} \frac{c_{38} \chi^3}{m^k \Lambda^3} \frac{\Lambda^2 \mathbf{P}(Y_0 = 0)^2}{(\log n)^2} = \frac{c_{38} \chi^3}{2} \frac{m_{k+\ell} \eta \Lambda^2 \mathbf{P}(Y_0 = 0)}{(\log n)^2}.
\]

\(^6\)The choice of $\frac{1}{d(m-1)}$ is to ensure $R := \frac{\Lambda}{n \chi \mathbf{P}(Y_0 = 0)} \geq 3$.

\(^7\)Theorem 4.2 only requires $n \geq 0$; we take $n \geq 3$ so that $\log n \geq \log 3 > 1$. 


date: May 5, 2020
Since \( \lambda_n^2 = \min\{\Lambda^2, n^{2\beta}\} \geq \frac{1}{(m-1)^2} \) (recalling that \( \Lambda \geq \frac{1}{m-1} \) for critical systems), this yields, with \( c_{42} := \frac{c_{43} m^\ell}{2(m-1)^2} \),

\[
\mathbb{E}(X_{n+k+\ell}) \geq c_{42} \frac{m^\ell \eta P(Y_0 = 0)}{\Lambda^3 (\log n)^2} \geq c_{42} \frac{m^\ell (\eta \chi P(Y_0 = 0))^3}{(\log n)^2} = c_{42} \frac{R^3 (\log n)^2}{R^3 (\log n)^2}.
\]

We take \( \ell := [c_{43} (\log R + \log \log n)] \), where \( c_{43} > 0 \) is a constant depending only on \( m \) and \( \beta \), such that \( c_{42} m^\ell \geq 2 \). On the other hand, the condition \( \ell \leq \frac{c_{43} m^2}{2\Lambda^2} \eta \) is respected if we take \( n := [c_{44} (\log R)^{1/2}] \) for some appropriate constant \( c_{44} \geq 1 \) whose value depends only on \( m \) and \( \beta \); moreover, \( c_{44} \geq 1 \) is taken to be sufficiently large so that \( \log R \leq n \). Accordingly,

\[
\mathbb{E}(X_{n+k+\ell}) \geq 2.
\]

By definition, \( n \leq \frac{n \Lambda}{\lambda_n} \leq \chi^{-3/2} \frac{n \Lambda \log n}{\lambda_n} \); also, \( \log R \leq n \) and \( \log \log n \leq n \), thus \( \log R + \log \log n \leq 2n \leq 2 \chi^{-3/2} \frac{n \Lambda \log n}{\lambda_n} \), which yields \( \ell := [c_{43} (\log R + \log \log n)] \leq 2c_{43} \chi^{-3/2} \frac{n \Lambda \log n}{\lambda_n} \). And of course, \( k < c_{36} \chi^{-3/2} \frac{n \Lambda \log n}{\lambda_n} \) by definition. As a result, with \( c_{45} := 1 + 2c_{43} + c_{36} \), we have

\[
n+k+\ell < \frac{c_{45} n \Lambda \log n}{\lambda_n [P(Y_0 = 0)]^{1/2}} = \frac{c_{45}}{[\chi^3 P(Y_0 = 0)]^{1/2}} \max\{\lambda n^{1-\beta}, n\} \log n,
\]

with \( n := [c_{44} (\log R)^{1/2}] \). This yields (6.6).

### 7. Proof of Theorem 1.1: lower bound

Let \((X_n, n \geq 0)\) be a system satisfying \( \mathbb{E}(X_0^3 m_X) < \infty \) and let \( p = p_c + \varepsilon \). Let \((Y_n, n \geq 0)\) be the critical system coupled with \((X_n, n \geq 0)\) as in the \( XY \)-coupling in Theorem 4.1. Then \( X_0 \geq Y_0 \) a.s., \( \mathbb{E}(X_0 - Y_0) = \varepsilon \mathbb{E}(X_0^*) \), \( \Lambda = \Lambda(Y_0) := \mathbb{E}(Y_0^3 m_Y) = p_c \mathbb{E}(X_0^3 m_X^3) < \infty \) and \( P(Y_0 = 0) = 1 - p_c \). By Lemma 2.4, \((Y_n, n \geq 0)\) is 0-regular with coefficient \( \chi := 1 - p_c \) in the sense of (2.3). It follows from Theorem 6.5 that there exists a constant \( c_{46} > 0 \) such that for all sufficiently small \( \varepsilon > 0 \), the free energy \( F_{\infty}(p_c + \varepsilon) \) of the system \((X_n, n \geq 0)\) satisfies

\[
F_{\infty}(p_c + \varepsilon) \geq \exp\left\{ - c_{46} \frac{[\log(\frac{1}{\varepsilon})]^{3/2}}{\varepsilon^{1/2}} \right\},
\]

which readily yields the lower bound in Theorem 1.1.
8. Proof of Theorems 1.2 and 1.3. We start with a simple comparison result, which is useful in the proof of the upper bound in Theorems 1.2 and 1.3.

**Lemma 8.1.** Let \((U_n, n \geq 0)\) and \((V_n, n \geq 0)\) be two recursive systems. If \(U_0 \geq V_0\) a.s. and if \(E(U_0) < \infty\), then for all \(n \geq 0\),

\[
0 \leq E(U_n) - E(V_n) \leq m^n [E(U_0) - E(V_0)].
\]

**Proof.** Since \(U_0 \geq V_0\) a.s., we can couple the two systems so that \(U_n \geq V_n\) a.s. for all \(n \geq 0\); in particular, \(E(U_n) - E(V_n) \geq 0\).

Let \((X_n, n \geq 0)\) be an arbitrary recursive system. Recall from the recurrence relation (3.4) that \(E(s^{X_{n+1}}) = \frac{1}{m} [E(s^{X_n})]^m + (1 - \frac{1}{m}) [P(X_n = 0)]^m\). Differentiating with respect to \(s\), and taking \(s = 1\), this yields \(E(X_{n+1}) = mE(X_n) - 1 + [P(X_n = 0)]^m\) for all \(n \geq 0\). Iterating the identity yields that for \(n \geq 0\),

\[
E(X_n) = m^n E(X_0) - \sum_{i=0}^{n-1} m^{n-i-1} \{1 - [P(X_i = 0)]^m\}.
\]

Applying the identity to \(X_n = U_n\) and to \(X_n = V_n\), and taking the difference, we obtain:

\[
E(U_n) - E(V_n) = m^n [E(U_0) - E(V_0)] + \sum_{i=0}^{n-1} m^{n-i-1} \{[P(U_i = 0)]^m - [P(V_i = 0)]^m\}.
\]

By the coupling, \(U_i \geq V_i\) a.s. so \([P(U_i = 0)]^m - [P(V_i = 0)]^m \leq 0\). The lemma follows. \(\square\)

Theorems 1.2 and 1.3 are proved by means of a truncation argument. Let \(X_0^*\) be a random variable taking values in \(\{1, 2, \ldots\}\) and satisfying

\[
P(X_0^* = k) \sim c_0 m^{-k} k^{-\alpha}, \quad k \to \infty,
\]

for some \(0 < c_0 < \infty\) and \(2 \leq \alpha \leq 4\) (in Theorem 1.2, we assume \(2 < \alpha \leq 4\), whereas in Theorem 1.3, we assume \(\alpha = 2\)). Let \(\ell_0 \geq 1\) be the smallest integer \(k \geq 1\) such that \(P(X_0 = k) > 0\). Let \(M > \ell_0\) be an integer (which will ultimately tend to infinity). Let \(\rho = \rho(M, \alpha)\) be a random variable
taking values in \( \{0, \ell_0\} \); we are going to give the distribution of \( \varrho \) later. Consider the truncated random variable

\[
X^{(M)}_0 := \varrho \mathbf{1}_{\{X^*_0 = \ell_0\}} + X^*_0 \mathbf{1}_{\{\ell_0 < X^*_0 \leq M\}}.
\]

Let \( Y^{(M)}_0 \) be a random variable whose distribution is given by

\[
P_{Y^{(M)}_0} = p_M P_{X^{(M)}_0} + (1 - p_M) \delta_0,
\]

where \( p_M \in (0, 1) \) is such that

\[
E_{p_M} \left( m Y^{(M)}_0 \right) = \left( m - 1 \right) E_{p_M} \left( m^{Y^{(M)}_0} \right).
\]

[The subscript \( p_M \) is to indicate the weight \( p_M \) of \( X^{(M)}_0 \) in the distribution of \( Y^{(M)}_0 \).] The existence of \( p_M \in (0, 1) \) is simple; actually the value of \( p_M \) can be explicitly computed:

\[
p_M = \frac{1}{1 + E\left( \left[ (m - 1)X^{(M)}_0 - 1 \right] m^{X^{(M)}_0} \right)} \in (0, 1).
\]

Hence

\[
\frac{1}{p_M} - 1 = E\left( \left[ (m - 1)X^{(M)}_0 - 1 \right] m^{X^{(M)}_0} \right)
\]

\[
= E\left( \left[ (m - 1)X^*_0 - 1 \right] m^{X^*_0} \mathbf{1}_{\{X^*_0 \leq M\}} \right) - P(X^*_0 > M) - a_M,
\]

where

\[
a_M := E\left( \left[ (m - 1)\ell_0 - 1 \right] m^{\ell_0} - \left[ (m - 1)\varrho - 1 \right] m^{\varrho} \right) \mathbf{1}_{\{X^*_0 = \ell_0\}}.
\]

In case \( 2 < \alpha \leq 4 \), we have \( p_c(X^*_0) > 0 \), and by (1.4),

\[
\frac{1}{p_c(X^*_0)} - 1 = E\left( \left[ (m - 1)X^*_0 - 1 \right] m^{X^*_0} \right),
\]

which, in view of (8.4), yields that for \( 2 < \alpha \leq 4 \),

\[
\frac{1}{p_c(X^*_0)} - \frac{1}{p_M} = P(X^*_0 > M) + a_M + E\left( \left[ (m - 1)X^*_0 - 1 \right] m^{X^*_0} \mathbf{1}_{\{X^*_0 > \ell_0\}} \right).
\]
8.1. Proof of Theorems 1.2 and 1.3: upper bound. To prove the upper bound, the choice of \( \varrho \) in (8.2) is simple: we choose \( \varrho = X_0^* \) (so the truncated random variable \( X_0^*(M) \) in (8.2) is simply \( X_0^* \mathbf{1}_{\{X_0^* \leq M\}} \)). As such, \( a_M = 0 \) in (8.5).

The next lemma gives the asymptotic behaviour of \( p_M - p_c(X_0^*) \) when \( M \to \infty \). Note that \( p_c(X_0^*) = 0 \) in case \( \alpha = 2 \).

**Lemma 8.2.** Assume \( \mathbb{P}(X_0^* = k) \sim c_0 m^{-k} k^{-\alpha}, k \to \infty, \) for some \( 0 < c_0 < \infty \) and \( 2 \leq \alpha \leq 4 \). When \( M \to \infty \),

\[
(8.7) \quad p_M - p_c(X_0^*) \sim \begin{cases} \frac{(m-1)c_0 p_c(X_0^*)^2}{\alpha-2} \frac{1}{M^{\alpha-2}}, & \text{if } 2 < \alpha \leq 4, \\ \frac{1}{(m-1)c_0} \frac{1}{\log M}, & \text{if } \alpha = 2. \end{cases}
\]

**Proof of Lemma 8.2.** We first treat the case \( 2 < \alpha \leq 4 \). Since \( a_M = 0 \), (8.6) becomes

\[
\frac{1}{p_c(X_0^*)} - \frac{1}{p_M} = \mathbb{P}(X_0^* > M) + \mathbb{E}\{[(m-1)X_0^* - 1]mX_0^* \mathbf{1}_{\{X_0^* > M\}}\};
\]

using the asymptotics of \( X_0^* \) given in (8.1), we obtain

\[
(8.8) \quad \frac{1}{p_c(X_0^*)} - \frac{1}{p_M} \sim \frac{(m-1)c_0}{\alpha-2} \frac{1}{M^{\alpha-2}}, \quad M \to \infty.
\]

This yields (8.7) in case \( 2 < \alpha \leq 4 \).

We now treat the case \( \alpha = 2 \). By the asymptotics of \( X_0^* \) given in (8.1), we have, in case \( \alpha = 2 \),

\[
\mathbb{E}\{[(m-1)X_0^* - 1]mX_0^* \mathbf{1}_{\{X_0^* \leq M\}}\} \sim (m-1)c_0 \log M, \quad M \to \infty.
\]

In view of (8.4) (and using the fact \( a_M = 0 \), as well as the trivial inequality \( \mathbb{P}(X_0^* > M) \leq 1 \)), this implies

\[
(8.9) \quad \frac{1}{p_M} \sim (m-1)c_0 \log M, \quad M \to \infty,
\]

which, in turn, yields (8.7) in case \( \alpha = 2 \). \( \square \)

We have now all the ingredients for the proof of the upper bound in Theorems 1.2 and 1.3.

Let \( (Y_n^{(M)}, n \geq 0) \) be a recursive system with initial distribution \( Y_0^{(M)} \). Then the system is critical; in fact, (8.3) is a rewriting of (1.4). By Theorem A (see the introduction), the free energy of the system is 0. It follows from the first inequality in (1.5) that \( \mathbb{E}_{p_M}(Y_n^{(M)}) \leq \frac{1}{m-1} \) for all \( n \geq 0 \).
Consider also the system \((X_n, n \geq 0)\) under \(P_{PM}\), i.e., with initial law \(P_{X_0} = p_M P_{X_0'} + (1 - p_M)\delta_0\). Since \(X_0 \geq Y_0^{(M)}\) \(P_{PM}\)-a.s., we are entitled to apply Lemma 8.1 to see that for all \(n \geq 0\),

\[
0 \leq E_{PM}(X_n) - E_{PM}(Y_n^{(M)}) \leq m^n [E_{PM}(X_0) - E_{PM}(Y_0^{(M)})].
\]  

(8.10) We take \(n = M\). On the right-hand side, note that \(E_{PM}(X_0) - E_{PM}(Y_0^{(M)}) = p_M E(X_0^* 1_{\{X_0^* > M\}}) \leq E(X_0^* 1_{\{X_0^* > M\}}),\) so

\[
m^M [E_{PM}(X_0) - E_{PM}(Y_0^{(M)})] \leq m^M \sum_{k=M+1}^{\infty} k P(X_0^* = k),
\]

which tends to 0 as \(M \to \infty\) (by (8.1)). Consequently, (8.10) yields that for all \(M\) sufficiently large (say \(M \geq M_0\)), we have \(0 \leq E_{PM}(X_M) - E_{PM}(Y_M^{(M)}) \leq 1\). We have already seen that \(E_{PM}(Y_M^{(M)}) \leq \frac{1}{m-1}\). Hence \(E_{PM}(X_M) \leq 1 + \frac{1}{m-1} = \frac{m}{m-1}\) (for \(M \geq M_0\)). By the second inequality in (1.5), we obtain that the free energy of the system \((X_n, n \geq 0)\) with \(p = p_M\) satisfies

(8.11) \[F_{\infty}(p_M) \leq \frac{E_{PM}(X_M)}{m^M} \leq \frac{m}{m-1} e^{-M \log m}, \quad \forall M \geq M_0.\]

Let \(p \in [p_{M+1}, p_M]\). When \(p\) is sufficiently close to \(p_c(X_0^*)\) (which is equivalent to saying that \(M\) is sufficiently large), we have, by (8.11),

(8.12) \[F_{\infty}(p) \leq F_{\infty}(p_M) \leq \frac{m}{m-1} e^{-M \log m}.\]

In case \(2 < \alpha \leq 4\), we deduce from Lemma 8.2 that \(\frac{m}{m-1} e^{-M \log m} = \exp(-\frac{c_{47} + o(1)}{(p_M - p_c(X_0^*))^{1/(\alpha - 2)}})\), where \(c_{47} := \frac{(m-1) p_c(X_0^*)^2}{\alpha - 2}\log m\). Note that as \(p \downarrow p_c(X_0^*)\), \(\frac{1}{(p_M - p_c(X_0^*))^{1/(\alpha - 2)}} = \frac{1 + o(1)}{(p_M - p_c(X_0^*))^{1/(\alpha - 2)}} \geq \frac{1+o(1)}{(p - p_c(X_0^*))^{1/(\alpha - 2)}}\). Hence

(8.13) \[F_{\infty}(p) \leq \exp \left( -\frac{c_{47} + o(1)}{(p - p_c(X_0^*))^{1/(\alpha - 2)}} \right), \quad p \downarrow p_c(X_0^*).\]

This yields the upper bound in Theorem 1.2.\footnote{The proof of (8.13) does not require \(\alpha \leq 4\). In other words, (8.13) holds for all \(\alpha > 2\); however, the exponent \(\frac{1}{\alpha - 2}\) is not optimal when \(\alpha > 4\), as we have seen in Theorem 1.1.}

In case \(\alpha = 2\), Lemma 8.2 says that \(\frac{m}{m-1} e^{-M \log m} = \exp\left(-\frac{e^{(1+o(1))}/(m-1)p_0)}{(p - p_c(X_0^*))^{1/(\alpha - 2)}}\right)\), \(p \downarrow 0\). The upper bound in Theorem 1.3 follows immediately from (8.12). \(\square\)
8.2. Proof of Theorems 1.2 and 1.3: lower bound. To prove the lower bound, here is our choice of \( \varrho \) in (8.2): for \( \alpha = 2 \), \( \varrho \) is simply 0; for \( 2 < \alpha \leq 4 \), \( \varrho \) is chosen to be independent of \( X_0^* \), with \( \mathbb{P}(\varrho = 0) = \frac{1}{M^{\alpha - 2}} = 1 - \mathbb{P}(\varrho = \ell_0) \).

**Lemma 8.3.** Assume \( \mathbb{P}(X_0^* = k) \sim c_0 m^{-k} k^{-\alpha} \), \( k \to \infty \), for some \( 0 < c_0 < \infty \) and \( 2 \leq \alpha \leq 4 \). When \( M \to \infty \),

\[
    p_M - p_c(X_0^*) \sim \begin{cases} 
        \frac{c_{48}}{M^{\alpha - 2}}, & \text{if } 2 < \alpha \leq 4, \\
        \frac{1}{(m-1)c_0} \log M, & \text{if } \alpha = 2,
    \end{cases}
\]

where, in case \( 2 < \alpha \leq 4 \), \( c_{48} := [p_c(X_0^*)]^2 c_{49} > 0 \), with

\[
    c_{49} := \frac{(m-1)c_0}{\alpha - 2} + \{(m-1)\ell_0 - 1\} \mathbb{P}(X_0^* = \ell_0) = \frac{c_{50}}{M^{\alpha - 2}},
\]

**Proof of Lemma 8.3.** The proof is as in the proof of Lemma 8.2, except that this time, we need to evaluate \( a_M \) with our new choice of \( \varrho \). Let us only point out the difference.

In case \( 2 < \alpha \leq 4 \), we have, by definition of \( a_M \) in (8.5),

\[
    a_M = \mathbb{E}\{[(m-1)\ell_0 - 1]\mathbb{P}(X_0^* = \ell_0) \mathbb{P}(X_0^* = \ell_0) = \frac{c_{50}}{M^{\alpha - 2}},
\]

where \( c_{50} := \{(m-1)\ell_0 - 1\} \mathbb{P}(X_0^* = \ell_0) > 0 \). Then (8.8) in the proof of Lemma 8.2 becomes

\[
    \frac{1}{p_c(X_0^*)} - \frac{1}{p_M} \sim \frac{(m-1)c_0}{\alpha - 2} \frac{1}{M^{\alpha - 2}} + a_M = \frac{c_{49}}{M^{\alpha - 2}}, \quad M \to \infty,
\]

with \( c_{49} > 0 \) being in (8.15). This yields Lemma 8.3 in case \( 2 < \alpha \leq 4 \).

Assume now \( \alpha = 2 \) (so \( p_c(X_0^*) = 0 \)). We have chosen \( \varrho = 0 \) in this case. By definition of \( a_M \) in (8.5),

\[
    a_M = \{(m-1)\ell_0 - 1\} \mathbb{P}(X_0^* = \ell_0) = c_{50} > 0.
\]

Then (8.9) in the proof of Lemma 8.2 becomes

\[
    \frac{1}{p_M} \sim (m-1)c_0 (\log M) - c_{50} \sim (m-1)c_0 \log M, \quad M \to \infty.
\]

This yields Lemma 8.3 in case \( \alpha = 2 \). \( \square \)
Recall from (2.3) that a system with initial value $X_0$ is said to be $\beta$-regular with coefficient $\chi$ if for all integers $j \geq 1$,

$$\Xi_j(X_0) := \mathbb{E}[(X_0 \land j)^2 \left( (m - 1)X_0 - 1 \right)] \geq \chi \min\{\mathbb{E}(X_0^3 m^{X_0}), j^\beta \}.$$

**Lemma 8.4.** Let $\alpha \in [2, 4]$. There exists a constant $c_{51} \in (0, 1]$ whose value depends only on $m$, $\alpha$, and the distribution of $X_0^*$, such that for all sufficiently large $M$, the critical system $(Y_{\alpha}^{(M)}, n \geq 0)$ is $(4 - \alpha)$-regular with coefficient $c_{51}$.

**Proof.** When $\alpha = 4$, Lemma 2.4 says that $(Y_{n}^{(M)}, n \geq 0)$ is 0-regular with coefficient $1 - p_M$, which goes to $1 - p_c(X_0) > 0$ when $M \to \infty$, so the result follows immediately.

Let $\alpha \in [2, 4]$ now. By assumption (8.1), $P(X_0^* = k) \sim c_0 m^{-k} k^{-\alpha}$, $k \to \infty$, so there exist constants $c_{52} > 0$ and $c_{53} > 0$, and an integer $k_0 \geq 2$ (for further use, we take $k_0$ to satisfy also $k_0 \geq \frac{2}{\alpha}$), all depending only on $m, \alpha$, and the law of $X_0^*$, such that for $M \geq k_0$,

\begin{align}
(8.16) & \quad \mathbb{P}(Y_{\alpha}^{(M)} = k) \leq c_{52} m^{-k} k^{-\alpha}, \quad \forall 1 \leq k \leq M, \\
(8.17) & \quad \mathbb{P}(Y_{\alpha}^{(M)} = k) \geq c_{53} m^{-k} k^{-\alpha}, \quad \forall k_0 \leq k \leq M.
\end{align}

So for all integer $M \geq 2k_0$ and real number $b \in [k_0, \frac{M}{2}]$, we have, by (8.17),

$$\mathbb{E}[m^{Y_{\alpha}^{(M)} - Y_{0}^{(M)}} \mathbf{1}_{\{b \leq Y_{\alpha}^{(M)} \leq 2b\}}] \geq \sum_{k \in [b, 2b] \cap \mathbb{Z}} m^k k c_{53} m^{-k} k^{-\alpha}$$

$$\geq c_{53} \sum_{k \in [b, 2b] \cap \mathbb{Z}} (2b)^{1-\alpha}$$

$$\geq c_{54} b^{2-\alpha},$$

with $c_{54} := 2^{-\alpha} c_{53}$. Furthermore, recalling that we only care about the case $\alpha \in [2, 4]$, we have, by (8.16),

$$\mathbb{E}(Y_{\alpha}^{(M)} m^{Y_{0}^{(M)}} \leq \sum_{k=1}^{M} k^3 m^k c_{52} m^{-k} k^{-\alpha} \leq c_{55} M^{4-\alpha} ,$$

where $c_{55} := \frac{2^{4-\alpha} c_{52}}{4-\alpha}$ (because $\sum_{k=1}^{M} k^{3-\alpha} \leq \int_{0}^{M+1} x^{3-\alpha} dx = \frac{(M+1)^{4-\alpha}}{4-\alpha} = \frac{M^{1-\alpha}}{4-\alpha} (1 + \frac{1}{M})^{4-\alpha} \leq \frac{M^{1-\alpha}}{4-\alpha} 2^{4-\alpha}$).
Let \( j \geq 1 \) be an integer. We distinguish two possible situations.

**First situation:** \( M \geq 2k_0j \). We use

\[
\Xi_j(Y_0^{(M)}) := \mathbf{E}[m_0^{Y_0^{(M)}}((m-1)Y_0^{(M)} - 1)(Y_0^{(M)} \wedge j)^2]
\]

\[
\geq \mathbf{E}[m_0^{Y_0^{(M)}}((m-1)Y_0^{(M)} - 1)(Y_0^{(M)} \wedge j)^2 \mathbf{1}_{\{k_0j \leq Y_0^{(M)} \leq 2k_0j\}}].
\]

On the event \( \{k_0j \leq Y_0^{(M)} \leq 2k_0j\} \) (so \( Y_0^{(M)} \geq \frac{2}{m-1} \)), we have \((m-1)Y_0^{(M)} - 1 \geq \frac{m-1}{2}Y_0^{(M)} \) and \( Y_0^{(M)} \wedge j = j \), so

\[
\Xi_j(Y_0^{(M)}) \geq \frac{m-1}{2} j^2 \mathbf{E}[m_0^{Y_0^{(M)}}Y_0^{(M)} \mathbf{1}_{\{k_0j \leq Y_0^{(M)} \leq 2k_0j\}}] \geq \frac{m-1}{2} c_{54} k_0^{2-\alpha} j^{4-\alpha},
\]

the last inequality being a consequence of (8.18).

**Second (and last) situation:** \( M < 2k_0j \) (and \( M \geq 2k_0 \)). This time, we start with

\[
\Xi_j(Y_0^{(M)}) \geq \mathbf{E}[m_0^{Y_0^{(M)}}((m-1)Y_0^{(M)} - 1)(Y_0^{(M)} \wedge j)^2 \mathbf{1}_{\{M \leq Y_0^{(M)} \leq M\}}].
\]

On the event \( \{M \leq Y_0^{(M)} \leq M\} \) (so \( Y_0^{(M)} \geq \frac{2}{m-1} \)), we have again \((m-1)Y_0^{(M)} - 1 \geq \frac{m-1}{2}Y_0^{(M)} \), whereas \( (Y_0^{(M)} \wedge j) \geq \frac{M}{2k_0} \). Consequently,

\[
\Xi_j(Y_0^{(M)}) \geq \frac{m-1}{2} \frac{M^2}{4k_0^2} \mathbf{E}[m_0^{Y_0^{(M)}}Y_0^{(M)} \mathbf{1}_{\{M \leq Y_0^{(M)} \leq M\}}] \geq \frac{m-1}{2} \frac{M^2}{4k_0^2} c_{54} \left(\frac{M}{2}\right)^{2-\alpha},
\]

the last inequality following again from (8.18). By (8.19), \( \mathbf{E}((Y_0^{(M)})^3 m_0^{Y_0^{(M)}}) \leq c_{55} M^{4-\alpha} \), so \( \Xi_j(Y_0^{(M)}) \geq c_{56} \mathbf{E}((Y_0^{(M)})^3 m_0^{Y_0^{(M)}}) \), with \( c_{56} := \frac{m-1}{2} \frac{c_{54}}{2^{4-\alpha} k_0^{2-\alpha} c_{55}} \).

In both situations, we see that for all \( M \geq 2k_0 \) and all integers \( j \geq 1 \),

\[
\Xi_j(Y_0^{(M)}) \geq c_{57} \min\{\mathbf{E}((Y_0^{(M)})^3 m_0^{Y_0^{(M)}}), j^{4-\alpha}\},
\]

with \( c_{57} := \min\{\frac{m-1}{2} c_{54} k_0^{2-\alpha}, c_{56}\} \). In other words, for all \( M \geq 2k_0 \), the system \((Y_0^{(M)}, n \geq 0)\) is \((4-\alpha)\)-regular with coefficient \( \min\{c_{57}, 1\} \). Lemma 8.4 is proved. \( \Box \)

We now proceed to the proof of the lower bound in Theorems 1.2 and 1.3. Again, we consider the system \((X_n, n \geq 0)\) under \( P_{PM} \), i.e., with initial
law $P_{X_0} = p_M P_{X_0} + (1 - p_M)\delta_0$, and couple it with the critical system $(Y_n^{(M)}, n \geq 0)$.

By definition, $P_{PM}(X_0 = k) \geq P(Y_0^{(M)} = k)$ for all integers $k \geq 1$. We intend to apply Theorem 6.5 to $(X_n, n \geq 0)$ under $P_{PM}$, so let us estimate $E[(Y_0^{(M)})^3 m_{Y_0^{(M)}}]_n$ and $E_{PM}(X_0) - E(Y_0^{(M)})$.

Since $X_0^{(M)} \leq X_0^\star 1_{\{X_0^\star \leq M\}}$, we have

$$E[(Y_0^{(M)})^3 m_{Y_0^{(M)}}] = p_M E[(X_0^{(M)})^3 m_{X_0^{(M)}}] \leq p_M E[(X_0^\star)^3 m_{X_0^\star} 1_{\{X_0^\star \leq M\}}].$$

By the tail behaviour of $X_0^\star$ given in (8.1), we know that for $M \to \infty$, $E[(X_0^\star)^3 m_{X_0^\star} 1_{\{X_0^\star \leq M\}}]$ is equivalent to $c_0 M^{4-\alpha}$ if $2 \leq \alpha < 4$, and to $c_0 \log M$ if $\alpha = 4$. So in both cases,

$$E[(Y_0^{(M)})^3 m_{Y_0^{(M)}}] \leq p_M M^{4-\alpha + o(1)}, \quad M \to \infty. \tag{8.20}$$

Let us now estimate $E_{PM}(X_0) - E(Y_0^{(M)})$. By definition,

$$E_{PM}(X_0) - E(Y_0^{(M)}) = p_M [E(X_0^\star) - E(X_0^{(M)})] \geq p_M [E(X_0^\star 1_{\{X_0^\star \leq M\}}) - E(X_0^{(M)})] = p_M [E(\ell_0 - \ell_0^\star) 1_{\{X_0^\star = \ell_0^\star\}}] = p_M [\ell_0 - E(\ell_0)] P(X_0^\star = \ell_0^\star).$$

For $2 < \alpha \leq 4$, $P(\ell_0 = 0) = \frac{1}{M^{1-\alpha}} = 1 - P(\ell_0 = \ell_0^\star)$, so $\ell_0 - E(\ell_0) = \frac{\ell_0}{M^{1-\alpha}}$. For $\alpha = 2$, $\ell_0 = 0 \Rightarrow \ell_0 - E(\ell_0) = \ell_0^\star$. In both cases, $\ell_0 - E(\ell_0) = \frac{\ell_0^\star}{M^{1-\alpha}}$, hence

$$E_{PM}(X_0) - E(Y_0^{(M)}) \geq \frac{\ell_0^\star P(X_0^\star = \ell_0^\star)}{M^{1-\alpha}} p_M. \tag{8.21}$$

Lemma 8.4 says the system $(Y_n^{(M)}, n \geq 0)$ is $(4-\alpha)$-regular in the sense of (2.3), with coefficient $\chi := c_{51}$. Now that (8.20) and (8.21) are established, and since $P(Y_0^{(M)} = 0) \geq 1 - p_M \geq \frac{1 - P(X_0^\star)}{2}$ (for all sufficiently large $M$), we are entitled to apply Theorem 6.5 to $(X_n, n \geq 0)$ under $P_{PM}$, to see that for $M \to \infty$, the free energy of $(X_n, n \geq 0)$ under $P_{PM}$ satisfies

$$F_\infty(p_M) \geq \exp \left( - M^{1+o(1)} [\log(M^{2+o(1)})]^{\max\{2, \alpha/2\}} \right) = \exp \left( - M^{1+o(1)} \right).$$

By Lemma 8.3, for $M \to \infty$, $M^{1+o(1)} = \frac{1}{(p_M - p_c(X_0^\star))^{1/(1-\alpha)} + \alpha M}$ if $2 < \alpha \leq 4$, and $M^{1+o(1)} = \frac{1}{(p_M - p_c(X_0^\star))^{1/(1-\alpha)} + \alpha M}$ if $\alpha = 2$. This yields the lower bound in Theorems 1.2 and 1.3 along the sequence $p = p_M$. The monotonicity of
$p \mapsto F_\infty(p)$ and the properties $\frac{p_M - p_c(X^*_p)}{p_{M+1} - p_c(X^*_p)} \to 1$ (if $2 < \alpha \leq 4$) and $\frac{p_M}{p_{M+1}} \to 1$ (if $\alpha = 2$; both properties are consequences of Lemma 8.3) allow us, exactly as in the proof of the upper bound in Theorems 1.2 and 1.3, to get rid of the restriction to the sequence $(p_M)$.

Acknowledgements. We are grateful to two anonymous referees, whose insightful comments have led to improvements in the presentation of the paper.

References.

[1] Aldous, D.J. and Bandyopadhyay, A. (2005). A survey of max-type recursive distributional equations. Ann. Appl. Probab. 15, 1047–1110.
[2] Alexander, K.S. (2008). The effect of disorder on polymer depinning transitions. Commun. Math. Phys., 279, 117–146.
[3] Berger, Q., Giacomin, G. and Lacoin, H. (2019). Disorder and critical phenomena: the $\alpha = 0$ copolymer model. Probab. Theory Related Fields 174, 787–819.
[4] Chen, X., Derrida, B., Hu, Y., Lifshits, M. and Shi, Z. (2019). A max-type recursive model: some properties and open questions. In: Sojourns in Probability Theory and Statistical Physics-III (pp. 166–186). Springer, Singapore.
[5] Collet, P., Eckmann, J.P., Glaser, V. and Martin, A. (1984). A spin glass with random couplings. J. Statist. Phys. 36, 89–106.
[6] Collet, P., Eckmann, J.P., Glaser, V. and Martin, A. (1984). Study of the iterations of a mapping associated to a spin-glass model. Commun. Math. Phys. 94, 353–370.
[7] Derrida, B., Hakim, V. and Vannimenus, J. (1992). Effect of disorder on two-dimensional wetting. J. Statist. Phys. 66, 1189–1213.
[8] Derrida, B. and Retaux, M. (2014). The depinning transition in presence of disorder: a toy model. J. Statist. Phys. 156, 268–290.
[9] Giacomin, G. (2007). Random Polymer Models. Imperial College Press.
[10] Giacomin, G. (2011). Disorder and critical phenomena through basic probability models. École d’été Saint-Flour XL (2010), Lecture Notes in Mathematics 2025, Springer, Heidelberg.
[11] Giacomin, G. and Toninelli, F.L. (2006). Smoothing effect of quenched disorder on polymer depinning transitions. Commun. Math. Phys. 266, 1–16.
[12] Giacomin, G., Toninelli, F. and Lacoin, H. (2010). Marginal relevance of disorder for pinning models. Commun. Pure Appl. Math. 63, 233–265.
[13] Goldschmidt, C. and Przykucki, M. (2019). Parking on a random tree. Combin. Probab. Comput. 28, 23–45.
[14] Hu, Y., Mallein, B. and Pain, M. (2020). An exactly solvable continuous-time Derrida–Retaux model. Commun. Math. Phys. 375, 605–651.
[15] Hu, Y. and Shi, Z. (2018). The free energy in the Derrida–Retaux recursive model. *J. Statist. Phys.* **172**, 718–741.

[16] Jordan, J. (2002). Almost sure convergence for iterated functions of independent random variables. *Ann. Appl. Probab.* **12**, 985–1000.

[17] Li, D. and Rogers, T.D. (1999). Asymptotic behavior for iterated functions of random variables. *Ann. Appl. Probab.* **9**, 1175–1201.

[18] Monthus, C. (2017). Strong disorder renewal approach to DNA denaturation and wetting: typical and large deviation properties of the free energy. *J. Statist. Mech. Theory Exper.* 2017(1), 013301.

[19] Tang, L.H. and Chaté, H. (2001). Rare-event induced binding transition of heteropolymers. *Phys. Rev. Letters* **86**(5), 830.

---

**School of Mathematical Sciences**
Shanghai Jiaotong University
200240 Shanghai, China
E-mail: chenxin@sjtu.edu.cn

**Collège de France, PSL**
11 place Marcelin Berthelot
F-75231 Paris Cedex 05, France
E-mail: derrida@lps.ens.fr

**St. Petersburg State University**
Department of Mathematics and Computer Sciences
Universitetskaya emb., 7-9, Russian Federation
E-mail: mikhail@lifshits.org

**Laboratoire de Physique**
École Normale Supérieure
PSL, CNRS, F-75005 Paris, France
E-mail: victor.dagard@lps.ens.fr

**LAGA, USP**
99 av. J-B Clément
F-93430 Villetaneuse, France
E-mail: yueyun@math.univ-paris13.fr

**St. Petersburg State University**
Sorbonne Université Paris VI
4 place Jussieu, F-75252 Paris Cedex 05
France
E-mail: zhan.shi@upmc.fr