# **ON ASYMPTOTICS OF THE BETA COALESCENTS**

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## Abstract

We show that the total number of collisions in the exchangeable coalescent process driven by the beta (1, b) measure converges in distribution to a 1-stable law, as the initial number of particles goes to  $\infty$ . The stable limit law is also shown for the total branch length of the coalescent tree. These results were known previously for the instance b = 1, which corresponds to the Bolthausen–Sznitman coalescent. The approach we take is based on estimating the quality of a renewal approximation to the coalescent in terms of a suitable Wasserstein distance. Application of the method to beta (a, b)-coalescents with 0 < a < 1 leads to a simplified derivation of the known (2 - a)-stable limit. We furthermore derive asymptotic expansions for the moments of the number of collisions and of the total branch length for the beta (1, b)-coalescent by exploiting the method of sequential approximations.

*Keywords:* Absorption time; asymptotic expansion; beta coalescent; coupling; number of collisions; total branch length; Wasserstein distance

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## 1. Introduction

Pitman [29] and Sagitov [30] introduced exchangeable coalescent processes with multiple collisions, also known as  $\Lambda$ -coalescents. A counting process associated with the  $\Lambda$ -coalescent is a Markov chain  $\Pi_n = (\Pi_n(t))_{t\geq 0}$  with right-continuous paths, which starts with *n* particles,  $\Pi_n(0) = n$ , and terminates when a sole particle remains. The particles merge according to the rule: for each  $t \geq 0$ , when the number of particles is  $\Pi_n(t) = m > 1$ , each *k* tuple of them merge into one particle at probability rate

$$\lambda_{m,k} = \int_0^1 x^k (1-x)^{m-k} x^{-2} \Lambda(\mathrm{d}x), \qquad 2 \le k \le m, \tag{1.1}$$

where  $\Lambda$  is a given finite measure on the unit interval. The merging of two or more particles is called a *collision*. With every collision,  $\Pi_n$  jumps to a smaller value. When  $\Lambda$  is a Dirac mass

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at 0, the  $\Lambda$ -coalescent is the classical Kingman coalescent [25], in which every pair of particles merges at the unit rate and only binary mergers are possible. Another eminent instance, known as the Bolthausen–Sznitman coalescent [6], appears when  $\Lambda$  is the Lebesgue measure on [0, 1].

The subclass of *beta coalescents* are the processes driven by some beta measure on [0, 1] with density

$$\frac{\Lambda(\mathrm{d}x)}{\mathrm{d}x} = (\mathrm{B}(a,b))^{-1} x^{a-1} (1-x)^{b-1}, \qquad a,b>0, \tag{1.2}$$

where  $B(\cdot, \cdot)$  denotes Euler's beta function. This class is amenable to analysis due to the fact that the transition rates (1.1) can be expressed in terms of  $B(\cdot, \cdot)$ . For this reason and due to multiple connections with Lévy processes and random trees, beta coalescents have been the subject of intensive research; see [2], [3], [5], [8], [9], [17], [20], and [29]. We refer the reader to [4] for a survey and further references.

In this paper we study beta coalescents with parameter  $0 < a \le 1$ . Specifically, we are interested in the total number of collisions  $X_n$  and the total branch length of the coalescent tree  $L_n$ . Note that  $X_n$  is equal to the total number of particles born through collisions, and  $L_n$  is the cumulative lifetime of all particles from the start of the process to its termination. The variable  $L_n$  is closely related to the number of segregating sites  $M_n$ , the connection being that, given  $L_n$ , the distribution of  $M_n$  is Poisson with mean  $rL_n$  for some fixed mutation rate r > 0.

A principal contribution of this paper is the proof of convergence in distribution to a 1-stable law for  $X_n$  and  $L_n$  as  $n \to \infty$ . As in much of the previous work (see, for instance, [13] and [21]), we use a renewal approximation to  $\Pi_n$ . A novel element in this context is estimating the quality of approximation in terms of a Wasserstein distance.

Our second new contribution is the derivation of asymptotic expansions for the moments of  $X_n$ ,  $L_n$ , and  $M_n$  for the beta (1, b)-coalescent with arbitrary parameter b > 0. These expansions are complementary to the results on convergence in distribution. The proofs of these asymptotic expansions are based on the method of sequential approximations, similarly as in [22].

The rest of the paper is organized as follows. In Section 2 we give a summary of some results on limit laws related to the beta coalescents. In Section 3 general properties of the block-counting Markov chain and basic recurrences are discussed, and the main results are stated. In Section 4 we recall the definition and properties of a Wasserstein distance. In Section 5 we provide proofs of the main results. Some auxiliary lemmas are collected in Appendix A.

## 2. A summary of limit laws for beta coalescents

In Tables 1–3 we summarize the limit laws for  $X_n$ ,  $L_n$ , and the absorption time  $\tau_n := \min\{t: \Pi_n(t) = 1\}$  of the coalescent. The distributions that appear in the tables are as follows.

- $\mathcal{N}$ , the standard normal distribution.
- $\delta_{\alpha}$  with  $1 < \alpha < 2$ , the (spectrally negative)  $\alpha$ -stable distribution with characteristic function

$$z \mapsto \exp\left\{|z|^{\alpha}\left(\cos\left(\frac{\pi\alpha}{2}\right) + i\sin\left(\frac{\pi\alpha}{2}\right)\operatorname{sgn}(z)\right)\right\}, \quad z \in \mathbb{R}.$$

•  $\delta_1$ , the (spectrally negative) 1-stable distribution with characteristic function

$$z \mapsto \exp\left\{-|z|\left(\frac{\pi}{2} - i\log|z|\operatorname{sgn}(z)\right)\right\}, \qquad z \in \mathbb{R}$$

- $\mathcal{E}_{\gamma}(a, b)$  with  $a, b, \gamma > 0$ , the law of the *exponential functional*  $\int_0^\infty e^{-\gamma S_{a,b}(t)} dt$ , where  $(S_{a,b}(t))_{t\geq 0}$  is a drift-free subordinator with Laplace exponent  $\Phi_{a,b}(z) = \int_0^1 (1-(1-x)^z) x^{a-3} (1-x)^{b-1} dx$ ,  $z \geq 0$ .
- $\mathcal{G}$ , the Gumbel with distribution function  $x \mapsto \exp\{-e^{-x}\}, x \in \mathbb{R}$ .
- $\rho$ , the convolution of infinitely many exponential laws with rates i(i-1)/2,  $i \ge 2$ .

а	b	$a_n$	$b_n$	Distribution	Source
0 < a < 1	<i>b</i> > 0	(1 - a)n	$(1-a)n^{1/(2-a)}$	\$2-a	This paper, [7], [13], and [21] $(b = 1)$
a = 1	<i>b</i> > 0	$\frac{n\log(n\log n)}{(\log n)^2}$	$\frac{n}{(\log n)^2}$	$\mathscr{S}_1$	This paper and [9], [20] $(b = 1)$
1 < a < 2	b > 0	0	$\frac{\Gamma(a)}{2-a} n^{2-a}$	$\mathcal{E}_{2-a}(a,b)$	[14] and [18]
a = 2	b > 0	$(2r_1)^{-1}(\log n)^2$	$(3^{-1}r_1^{-3}r_2\log^3 n)^{1/2}$	${\mathcal N}$	[14] and [22]
<i>a</i> > 2	b > 0	$m_1^{-1}\log n$	$(m_1^{-3}m_2\log n)^{1/2}$	$\mathcal{N}$	[14] and [15]

TABLE 1: Limit distributions of  $(X_n - a_n)/b_n$  for beta (a, b)-coalescents.

TABLE 2: Limit distributions of  $(\tau_n - a_n)/b_n$  for beta (a, b)-coalescents.

а	b	$a_n$	$b_n$	Distribution	Source
a = 0		0	1	ρ	[32]
a = 1	b = 1	$\log \log n$	1	Ģ	[17] and [11]
1 < a < 2	b > 0	$m^{-1}\log n$	$(m^{-3}s^2\log n)^{1/2}$	$\mathcal{N}$	[14]
a = 2	b > 0	$c_1^{-1}\log n$	$(c_1^{-3}c_2\log n)^{1/2}$	$\mathcal{N}$	[14]
<i>a</i> > 2	b > 0	$(\gamma m_1)^{-1}\log n$	$\gamma^{-1}(m_1^{-3}(m_2+m_1^2)\log n)^{1/2}$	$\mathcal{N}$	[14] and [15]

TABLE 3: Limit distributions of  $(L_n - a_n)/b_n$  for beta (a, b)-coalescents.

a	b	$a_n$	$b_n$	Distribution	Source
a = 0		$2\log n$	2	Ģ	[8] and [32]
$0 < a < \frac{3 - \sqrt{5}}{2}$	b = 2 - a	$c_1 n^a$	1	exists	[24]
$a = \frac{3 - \sqrt{5}}{2}$	b = 2 - a	$c_1 n^a$	$c_2(\log n)^{1/\alpha}$	$\$_{2-a}$	[24]
$\frac{3-\sqrt{5}}{2} < a < 1$	b = 2 - a	$c_1 n^a$	$c_2(\beta n^{-\beta})^{1/\alpha}$	\$2-a	[24]
a = 1	b > 0	$\frac{n\log(n\log n)}{b(\log n)^2}$	$\frac{n}{b(\log n)^2}$	$\mathscr{S}_1$	This paper and [8] $(b = 1)$ ,
a > 1	b > 0	0	B(a, b)n	$\mathcal{E}_1(a,b)$	[26], and [27]

In Table 1  $r_1 = \zeta(2, b)$  and  $r_2 = 2\zeta(3, b)$ , where  $\zeta(\cdot, \cdot)$  is the Hurwitz zeta function;  $m_1 = \Psi(a - 2 + b) - \Psi(b)$  and  $m_2 = \Psi'(b) - \Psi'(a - 2 + b)$ , where  $\Psi(\cdot)$  is the logarithmic derivative of the gamma function.

For the Bolthausen–Sznitman coalescent, the limit law of  $X_n$  was first obtained in [9] using singularity analysis of generating functions. A probabilistic proof of this result appeared in [20], where the coupling of a random walk with a barrier was exploited, and the technique was further extended in [21] to study collisions in the beta (a, 1)-coalescents with  $a \in (0, 2)$ . The aforementioned limit laws for a > 1 are specializations of results for more general  $\Lambda$ -coalescents with *dust component*, i.e. those driven by measures  $\Lambda$  such that  $\int_0^1 x^{-1} \Lambda(dx) < \infty$  [13]–[15], and [18]. For Kingman's coalescent, we have  $X_n = n - 1$  for all  $n \in \mathbb{N}$ .

In the next two tables the value a = 0 corresponds to Kingman's coalescent. In Table 2,

$$\begin{split} \mathbf{m} &= \frac{a+b-1}{(a-1)(2-a)} (1-(a+b-2)(\Psi(a+b-1)-\Psi(b))), \\ \mathbf{s}^2 &= \frac{a+b-1}{(a-1)(2-a)} \\ &\times (2(\Psi(a+b-1)-\Psi(b)) \\ &\quad -(a+b-2)((\Psi(a+b-1)-\Psi(b))^2+\Psi'(b)-\Psi'(a+b-1))), \end{split}$$

 $c_1 = b(b+1)\zeta(2, b)$ , and  $c_2 = 2b(b+1)\zeta(3, b)$ . The constants  $m_1$  and  $m_2$  are the same as in Table 1, and, for a > 2,  $\gamma = (a - 1 + b)(a - 2 + b)/((a - 1)(a - 2))$ .

For the case in which  $a \in (0, 1)$  and b > 0, the beta (a, b)-coalescent has the property of coming down from  $\infty$  [31], which implies that  $\tau_n$  weakly converges without any normalization to some limiting law, which is not known explicitly. The result for a > 1 is a special case of Theorem 4.3 of [14]. The case in which a = 1 and  $b \neq 1$  is open; in this case the coalescent does not come down from  $\infty$ .

In Table 3,  $\alpha = 2 - a$ ,  $\beta = 1 + \alpha - \alpha^2$ ,  $c_1 = \Gamma(\alpha + 1)(\alpha - 1)/(2 - \alpha)$ , and  $c_2 = \Gamma(\alpha + 1)(\alpha - 1)^{1 + \alpha^{-1}}/(\cos(\pi \alpha/2)\Gamma^{\alpha^{-1}}(2 - \alpha))$ .

In [26] the weak convergence of properly normalized  $L_n$  was proved for  $\Lambda$ -coalescents with a dust component. In particular, that result covered the beta (a, b)-coalescents with a > 1. Although some partial results for  $a \in (0, 1)$  and b > 0 were obtained in [7], this case with  $b \neq 2 - a$  remains open.

## 3. Main results

For the general  $\Lambda$ -coalescent, the Markov chain  $\Pi_n$  is a pure death process which jumps from state *m* to m - k + 1 at rate  $\binom{m}{k} \lambda_{m,k}$ , where  $\lambda_{m,k}$ ,  $2 \le k \le m$ , is given by (1.1). The total transition rate from state  $m \ge 2$  is

$$\lambda_m := \sum_{k=2}^m \binom{m}{k} \lambda_{m,k} = \int_0^1 (1 - mx(1 - x)^{m-1} - (1 - x)^m) x^{-2} \Lambda(\mathrm{d}x).$$
(3.1)

The first decrement  $I_n$  of  $\Pi_n$  has distribution

$$\mathbb{P}\{I_n = k\} = \binom{n}{k+1} \frac{\lambda_{n,k+1}}{\lambda_n}, \qquad 1 \le k \le n-1.$$

The strong Markov property of the coalescent entails the distributional recurrences

$$X_1 = 0, \qquad X_n \stackrel{\mathrm{D}}{=} 1 + X'_{n-I_n}, \qquad n \in \mathbb{N} \setminus \{1\},$$
 (3.2)

$$\tau_1 = 0, \qquad \tau_n \stackrel{\mathrm{D}}{=} T_n + \tau'_{n-I_n}, \qquad n \in \mathbb{N} \setminus \{1\},$$

$$L_1 = 0, \qquad L_n \stackrel{\mathrm{\tiny D}}{=} nT_n + L'_{n-I_n}, \qquad n \in \mathbb{N} \setminus \{1\}, \tag{3.3}$$

where  $T_n$  denotes the time of the first collision (hence,  $T_n$  has the exponential law with parameter  $\lambda_n$ ), and  $X'_k$  (respectively  $\tau'_k$  and  $L'_k$ ) is independent of  $I_n$  (are each independent of the pair  $(T_n, I_n)$ ) and is distributed as  $X_k$  (respectively  $\tau_k, L_k$ ) for each  $k \in \mathbb{N}$ .

Letting  $\Lambda$  be defined as in (1.2) with  $a \in (0, 1]$ , define

$$p_{n,k}^{(a)} := \mathbb{P}\{I_n = n - k\}, \qquad k = 1, \dots, n - 1.$$
 (3.4)

Using the leading terms of asymptotic relations (A.3), (A.4), and (A.5) below, we infer that

$$\lim_{n \to \infty} p_{n,n-k}^{(a)} = \frac{(2-a)\Gamma(k+a-1)}{\Gamma(a)(k+1)!} =: p_k^{(a)}, \qquad k \in \mathbb{N};$$

hence (see also [7, Lemma 2.1]),

$$I_n \xrightarrow{\mathrm{D}} \xi$$
 as  $n \to \infty$ ,

where  $\xi$  is a random variable with distribution  $(p_k^{(a)})_{k \in \mathbb{N}}$ .

Consider a zero-delayed random walk  $(S_n)_{n \in \mathbb{N}_0}$  defined by  $S_0 := 0$  and  $S_n := \xi_1 + \cdots + \xi_n$ for  $n \in \mathbb{N}$ , where  $\xi_1, \xi_2, \ldots$  are independent copies of  $\xi$  with distribution  $(p_k^{(a)})_{k \in \mathbb{N}}$ , and let  $(N_n)_{n \in \mathbb{N}_0}$  be the associated first passage time sequence defined by  $N_n := \inf\{k \ge 0: S_k \ge n\}$ ,  $n \in \mathbb{N}_0$ . It is plain that

$$N_0 = 0, \qquad N_n \stackrel{\mathrm{D}}{=} 1 + N'_{n-\xi \wedge n} = 1 + N'_{n-\xi} \,\mathbf{1}_{\{\xi < n\}}, \qquad n \in \mathbb{N}, \tag{3.5}$$

where  $N'_k$  is independent of  $\xi$  and distributed as  $N_k$  for each  $k \in \mathbb{N}$ . Comparing (3.2) and (3.5) we can expect that, if  $N_n$  (properly centered and normalized) converges weakly to some proper and nondegenerate probability law then the same is true for  $X_n$  (with the same centering and normalization). This is what we mean by a renewal approximation mentioned in the introduction. This idea was exploited in [13] for  $a \in (0, 1)$  and b > 0, and in [21] for  $a \in (0, 1]$  and b = 1 to derive the limit distribution of  $X_n$  from that of  $N_n$ . For  $a \in (0, 1]$  and b > 0, we will use a method based on probability metrics to show the stable limits.

**Theorem 3.1.** As  $n \to \infty$ , the number of collisions  $X_n$  in the beta (a, b)-coalescent satisfies

(i) for 0 < a < 1 and b > 0,

$$\frac{X_n - (1-a)n}{(1-a)n^{1/(2-a)}} \xrightarrow{\mathrm{D}} \mathscr{S}_{2-a},$$

(ii) for a = 1 and b > 0,

$$\frac{\log^2 n}{n} X_n - \log n - \log \log n \xrightarrow{\mathrm{D}} \mathscr{S}_1$$

As a consequence of our main theorem, we also obtain a weak limit for the total branch length  $L_n$  and the number of segregating sites  $M_n$  (see [26]) of the beta (1, b)-coalescent.

**Corollary 3.1.** For the total branch length  $L_n$  in the beta (1, b)-coalescent, we have, as  $n \to \infty$ ,

$$\frac{b\log^2 n}{n}L_n - \log n - \log\log n \xrightarrow{\mathrm{D}} \mathscr{S}_1.$$

**Corollary 3.2.** For the number of segregating sites  $M_n$  in the beta (1, b)-coalescent, we have, as  $n \to \infty$ ,

$$\frac{b\log^2 n}{rn}M_n - \log n - \log\log n \xrightarrow{\mathrm{D}} \mathscr{S}_1,$$

where r > 0 is the rate of the homogeneous Poisson process on branches of the coalescent tree.

We now turn to the moments of  $X_n$ ,  $L_n$ , and  $M_n$ . An analysis of these moments provides further insight into the structure of these functionals. Our next result concerns the asymptotics of the moments of the number of collisions  $X_n$  in the beta (1, b)-coalescent.

**Theorem 3.2.** Fix  $b \in (0, \infty)$  and  $j \in \mathbb{N}_0$ . The *j*th moment of the number of collisions  $X_n$  in the beta (1, b)-coalescent has the asymptotic expansion

$$\mathbb{E}X_n^j = \frac{n^j}{\log^j n} \left( 1 + \frac{m_j}{\log n} + O\left(\frac{1}{\log^2 n}\right) \right) \quad as \ n \to \infty, \tag{3.6}$$

where the sequence  $(m_j)_{j \in \mathbb{N}_0}$  is recursively defined via  $m_0 := 0$  and  $m_j := m_{j-1} + \kappa_j/j$  for  $j \in \mathbb{N}$ , with  $\kappa_j := (j + b - 1)\Psi(j + b) + j - (b - 1)\Psi(b), \ j \in \mathbb{N}_0$ .

For further information on the coefficients  $m_j$ ,  $j \in \mathbb{N}$ , we refer the reader to (5.5) below in the proof of the following corollary, which provides asymptotic expansions for the central moments of  $X_n$  in the beta (1, b)-coalescent.

**Corollary 3.3.** Fix  $b \in (0, \infty)$  and  $j \in \mathbb{N} \setminus \{1\}$ . The *j*th central moment of the number of collisions  $X_n$  in the beta (1, b)-coalescent has the asymptotic expansion

$$\mathbb{E}(X_n - \mathbb{E}X_n)^j = \frac{(-1)^j}{j} B(b, j-1) \frac{n^j}{\log^{j+1} n} + O\left(\frac{n^j}{\log^{j+2} n}\right) \quad as \ n \to \infty.$$
(3.7)

In particular,  $\operatorname{var}(X_n) = (2b)^{-1}n^2/\log^3 n + O(n^2/\log^4 n) \text{ as } n \to \infty.$ 

**Remark 3.1.** For b = 1, (3.7) reduces to the asymptotic expansion (see [28, p. 277 or Theorem 2.1 with  $\alpha = 0$ ])

$$\mathbb{E}(X_n - \mathbb{E}X_n)^j = \frac{(-1)^j}{j(j-1)} \frac{n^j}{\log^{j+1} n} + O\left(\frac{n^j}{\log^{j+2} n}\right) \quad \text{as } n \to \infty.$$

The last result concerns the moments and central moments of the total branch length  $L_n$  of the beta (1, b)-coalescent.

**Proposition 3.1.** *Fix*  $b \in (0, \infty)$  *and*  $j \in \mathbb{N}_0$ *. The jth moment of the total branch length*  $L_n$  *of the* beta (1, b)*-coalescent has the asymptotic expansion* 

$$\mathbb{E}L_n^j = \frac{1}{b^j} \frac{n^j}{\log^j n} \left( 1 + \frac{m_j}{\log n} + O\left(\frac{1}{\log^2 n}\right) \right) \quad as \ n \to \infty.$$

where the sequence  $(m_j)_{j \in \mathbb{N}_0}$  is defined as in Theorem 3.2. Moreover, for  $j \in \{2, 3, \ldots\}$ , the

*jth central moment of*  $L_n$  *has the asymptotic expansion* 

$$\mathbb{E}(L_n - \mathbb{E}L_n)^j = \frac{(-1)^j}{jb^j} \mathbf{B}(b, j-1) \frac{n^j}{\log^{j+1} n} + O\left(\frac{n^j}{\log^{j+2} n}\right) \quad as \ n \to \infty.$$

In particular,  $\operatorname{var}(L_n) = (2b^3)^{-1}n^2/\log^3 n + O(n^2/\log^4 n) \text{ as } n \to \infty.$ 

Proposition 3.1 indicates that  $bL_n$  essentially behaves as  $X_n$ , in agreement with the comparison of Theorem 3.1(ii) and Corollary 3.2. The proof of Proposition 3.1 essentially follows the same lines as the analogous proofs of Theorem 3.2 and Corollary 3.3 for  $X_n$ . Instead of the distributional recurrence (3.2) for  $(X_n)_{n \in \mathbb{N}}$ , we have to work with the distributional recurrence (3.3) for  $(L_n)_{n \in \mathbb{N}}$ . Since the expansion of  $\mathbb{E}T_n = 1/\lambda_n$  is known (see Lemma A.4), the proofs concerning  $X_n$  are readily adapted for  $L_n$ . A proof of Proposition 3.1 is therefore omitted. We finally mention that, for the beta (1, b)-coalescent with mutation rate r > 0, expansions for the moments and central moments of the number of segregating sites  $M_n$  can be easily obtained, since (see, for example, [8, p. 1417]) the descending factorial moments of  $M_n$  are related to the moments of  $L_n$  via  $\mathbb{E}(M_n)_j = r^j \mathbb{E}L_n^j$ ,  $j \in \mathbb{N}_0$ , where  $(M_n)_0 := 1$  and  $(M_n)_j := M_n(M_n - 1) \cdots (M_n - j + 1)$  for  $j \in \mathbb{N}$ .

# 4. Probability distances $\chi_T$ and $d_q$

For T > 0, the  $\chi_T$ -distance of two real-valued random variables X and Y is defined by

$$\chi_T(X, Y) = \sup_{|t| \le T} |\mathbb{E} e^{itX} - \mathbb{E} e^{itY}|.$$

By the continuity theorem for the characteristic functions, convergence in distribution  $Z_n \xrightarrow{D} Z$ holds if and only if  $\lim_{n\to\infty} \chi_T(Z_n, Z) = 0$  for every T > 0.

Let  $\mathcal{D}_q$ ,  $q \in (0, 1]$ , be the set of probability laws on  $\mathbb{R}$  with finite qth absolute moment. Recall that  $|x - y|^q$  is a metric on  $\mathbb{R}$ . The Wasserstein distance on  $\mathcal{D}_q$  is defined by

$$d_q(X,Y) = \inf \mathbb{E}|\widehat{X} - \widehat{Y}|^q, \qquad (4.1)$$

where the infimum is taken over all couplings  $(\widehat{X}, \widehat{Y})$  such that  $X \stackrel{\text{D}}{=} \widehat{X}$  and  $Y \stackrel{\text{D}}{=} \widehat{Y}$ .

For ease of reference, we summarize the properties of  $d_q$  in the following proposition.

**Proposition 4.1.** Let X and Y be random variables with finite qth absolute moments. The Wasserstein distance  $d_q$  has the following properties.

- (Dist)  $d_q(X, Y)$  depends only on the marginal distributions of X and Y.
- (Inf) The infimum in (4.1) is attained for some coupling.
- (Rep) The Kantorovich–Rubinstein representation holds, i.e.

$$d_q(X, Y) = \sup_{f \in \mathcal{F}_q} |\mathbb{E}f(X) - \mathbb{E}f(Y)|,$$

where  $\mathcal{F}_q := \{ f \in C(\mathbb{R}) : |f(x) - f(y)| \le |x - y|^q, x, y \in \mathbb{R} \}.$ 

- (Hom)  $d_q(cX, cY) = |c|^q d(X, Y)$  for  $c \in \mathbb{R}$ .
- (Reg) For X, Y, and Z defined on the same probability space,  $d_q(X + Z, Y + Z) \le d_q(X, Y)$ provided  $Z \in \mathcal{D}_q$  is independent of (X, Y).

(Aff) 
$$d_a(X + a, Y + a) = d_a(X, Y)$$
 for  $a \in \mathbb{R}$ .

(Conv) For  $X, X_n \in \mathcal{D}_q$ , the convergence  $d_q(X_n, X) \to 0$  as  $n \to \infty$  implies that  $X_n \xrightarrow{\mathrm{D}} X$ and  $\mathbb{E}|X_n|^q \to \mathbb{E}|X|^q$ .

*Proof.* We refer the reader to [12] and [23] for most of these facts. To prove (Reg), choose an independent of Z coupling (X', Y') on which the infimum in the definition of  $d_q$  is attained. Then  $X + Z \stackrel{\text{D}}{=} X' + Z$ ,  $Y + Z \stackrel{\text{D}}{=} Y' + Z$ , and the definition of  $d_q$  entails

$$d_q(X+Z,Y+Z) \le \mathbb{E}|(X'+Z) - (Y'+Z)|^q = \mathbb{E}|X'-Y'|^q = d_q(X,Y).$$

Property (Conv): the convergence of moments is easy; the rest is a consequence of Lemma 4.1 to follow.

**Lemma 4.1.** For T > 0 and  $q \in (0, 1]$ , there exists a constant  $C = C_{T,q} > 0$  such that  $\sup_{|t| \le T} |\mathbb{E}e^{itX} - \mathbb{E}e^{itY}| \le Cd_q(X, Y).$ 

*Proof.* Assume that the infimum in the definition of  $d_q(X, Y)$  is attained on  $(\hat{X}, \hat{Y})$ . It is easy to check that  $|e^{ix} - e^{iy}| = 2|\sin(x - y)/2| \le 2^{1-q}M_q|x - y|^q$  for  $q \in (0, 1]$  and  $x, y \in \mathbb{R}$ , where  $M_q := \sup_{u>0} |\sin u|u^{-q} < \infty$ . Hence,

$$\sup_{|t| \le T} |\mathbb{E}e^{itX} - \mathbb{E}e^{itY}| = \sup_{|t| \le T} |\mathbb{E}e^{itX} - \mathbb{E}e^{itY}|$$

$$\leq \sup_{|t| \le T} \mathbb{E}|e^{it\hat{X}} - e^{it\hat{Y}}|$$

$$\leq 2^{1-q}M_q \sup_{|t| \le T} |t|^q \mathbb{E}|\hat{X} - \hat{Y}|^q$$

$$= 2^{1-q}M_q T^q d_q(X, Y),$$

as required.

# 5. Proofs

## 5.1. Proof of Theorem 3.1

Suppose that a = 1. Set  $a_n := n^{-1} \log^2 n$  and  $b_n := \log n + \log \log n$  for  $n \in \mathbb{N}$ . It suffices to show that  $\lim_{n\to\infty} \chi_T(a_n X_n - b_n, \mathfrak{Z}_1) = 0$  for every T > 0. The triangle inequality yields

$$\chi_T(a_n X_n - b_n, \, \$_1) \le \chi_T(a_n X_n - b_n, \, a_n N_n - b_n) + \chi_T(a_n N_n - b_n, \, \$_1).$$

The second term converges to 0 by Proposition 2 of [20] on the stable limit for the number of renewals. In view of Lemma 4.1, to prove convergence to 0 of the first term, it suffices to check that  $\lim_{n\to\infty} d_q(a_nX_n - b_n, a_nN_n - b_n) = 0$  for some  $q \in (0, 1]$ , which in view of the properties (Hom) and (Aff) in Proposition 4.1 amounts to the estimate

$$d_q(X_n, N_n) = o(n^q \log^{-2q} n) \quad \text{as } n \to \infty.$$
(5.1)

Now assume that  $a \in (0, 1)$ . By Theorem 7 of [10] we have  $\chi_T(a_n N_n - b_n, \vartheta_{2-a}) \to 0$  for every T > 0 with  $a_n := (1 - a)^{-1} n^{-1/(2-a)}$  and  $b_n := n^{(1-a)/(2-a)}$ . By the same reasoning as above, proving Theorem 3.1 for  $a \in (0, 1)$  reduces to showing that

$$d_q(X_n, N_n) = o(n^{q/(2-a)}) \quad \text{as } n \to \infty, \tag{5.2}$$

for some  $q \in (0, 1]$ .

Using recurrences (3.2) for  $X_n$  and (3.5) for  $N_n$ , we obtain

$$t_{n} := d_{q}(X_{n}, N_{n})$$

$$= d_{q}(X'_{n-I_{n}}, N'_{n-(\xi \wedge n)})$$

$$\leq d_{q}(N'_{n-I_{n}}, N'_{n-(\xi \wedge n)}) + d_{q}(X'_{n-I_{n}}, N'_{n-I_{n}})$$

$$\leq d_{q}(N'_{n-I_{n}}, N'_{n-(\xi \wedge n)}) + \mathbb{E}|\widehat{X}_{n-I_{n}} - \widehat{N}_{n-I_{n}}|^{q}$$

$$=: c_{n} + \sum_{k=1}^{n-1} \mathbb{P}\{I_{n} = n - k\}\mathbb{E}|\widehat{X}_{k} - \widehat{N}_{k}|^{q}$$

for arbitrary pairs  $((\widehat{X}_k, \widehat{N}_k))_{1 \le k \le n-1}$  independent of  $I_n$  such that  $\widehat{X}_k \stackrel{\text{D}}{=} X_k$  and  $\widehat{N}_k \stackrel{\text{D}}{=} N_k$ . Passing to the infimum over all such pairs leads to

$$t_n \le c_n + \sum_{k=1}^{n-1} \mathbb{P}\{I_n = n - k\}t_k.$$
 (5.3)

We will use (5.3) to estimate  $t_n$ .

First we find an appropriate bound for  $c_n$ . Let  $(\hat{I}_n, \hat{\xi})$  be a coupling of  $I_n$  and  $\xi$  such that (recall property (Inf) of Proposition 4.1)  $d_q(I_n, \xi \wedge n) = \mathbb{E}|\hat{I}_n - \hat{\xi} \wedge n|^q$ . Let  $(\hat{N}_k)_{k \in \mathbb{N}}$  be a copy of  $(N_k)_{k \in \mathbb{N}}$  independent of  $(\hat{I}_n, \hat{\xi})$ . Since  $(\hat{I}_n, \hat{\xi}, (\hat{N}_k))$  is a particular coupling, we have  $c_n = d_q(N'_{n-I_n}, N'_{n-(\xi \wedge n)}) \leq \mathbb{E}|\hat{N}_{n-\hat{I}_n} - \hat{N}_{n-(\hat{\xi} \wedge n)}|^q$ . Using the stochastic inequality  $N_{x+y} - N_x \leq N_y, x, y \in \mathbb{N}$ , yields  $\mathbb{E}|\hat{N}_{n-\hat{I}_n} - \hat{N}_{n-\hat{\xi} \wedge n}|^q \leq \mathbb{E}\hat{N}_{|\hat{I}_n - \hat{\xi} \wedge n|}^q$ . Furthermore, we obviously have  $N_n \leq n$ ; hence,  $c_n \leq \mathbb{E}|\hat{I}_n - \hat{\xi} \wedge n|^q = d_q(I_n, \xi \wedge n)$ . Now we invoke the Kantorovich–Rubinstein representation, property (Rep) of Proposition 4.1, for  $d_q$ . Set  $\mathcal{F}_{q,0} := \mathcal{F}_q \cap \{f: f(0) = 0\}$ , and note that  $f \in \mathcal{F}_{q,0}$  implies that  $|f(x)| \leq |x|^q, x \in \mathbb{R}$ . We have

$$\begin{split} c_n &\leq d_q(I_n, \xi \wedge n) \\ &= \sup_{f \in \mathcal{F}_q} |\mathbb{E}f(I_n) - \mathbb{E}f(\xi \wedge n)| \\ &= \sup_{f \in \mathcal{F}_{q,0}} |\mathbb{E}f(I_n) - \mathbb{E}f(\xi \wedge n)| \\ &= \sup_{f \in \mathcal{F}_{q,0}} \left| \sum_{k=1}^{n-1} \mathbb{P}\{I_n = k\} f(k) - \sum_{k=1}^{n-1} \mathbb{P}\{\xi = k\} f(k) - f(n) \sum_{k \geq n} \mathbb{P}\{\xi = k\} \\ &\leq \sum_{k=1}^{n-1} |\mathbb{P}\{I_n = k\} - \mathbb{P}\{\xi = k\} | k^q + n^q \mathbb{P}\{\xi \geq n\}. \end{split}$$

For appropriate  $q \in (0, 1]$  (to be specified below) such that a + q > 1, use Lemma A.3 in Appendix A along with the relation  $\mathbb{P}\{\xi \ge n\} = O(n^{a-2})$  to obtain the estimate  $c_n = O(n^{q+a-2})$ . With this bound for  $c_n$ , a *O*-estimate for  $t_n$  follows using Lemma A.1.

If  $a \in (0, 1)$ , we can take q = 1. Then Lemma A.1 applies with  $\psi_n = n$  and  $r_n = Mn^{a-1}$  (large enough M) and gives the estimate  $d_q(X_n, N_n) = O(n^a)$ , which implies (5.2).

For the case a = 1, application of the same lemma with  $\psi_n = n/(\log(n + 1))$  and  $r_n = Mn^{q-1}$  (large enough M) leads to  $t_n \leq Mn^q (\log n)^{-1}$ . Thus, (5.1) holds for  $q \in (0, \frac{1}{2})$ . The proof is complete.

# 5.2. Proof of Corollaries 3.1 and 3.2

We follow closely the proofs of Theorem 5.2 and Corollary 6.2 of [8]. In view of

$$\frac{b\log^2 n}{n}L_n - \log n - \log\log n = \frac{\log^2 n}{n}X_n - \log n - \log\log n + \frac{\log^2 n}{n}(bL_n - X_n),$$

it is enough to show that  $((\log^2 n)/n)(bL_n - X_n) \rightarrow 0$  in  $L_2$ .

Let the  $T_j$  be independent exponential variables with rates  $\lambda_j$ ,  $j \ge 2$ . Assuming that the  $T_j$  are independent of the sequence of states visited by  $\Pi_n$ , we may identify  $T_j$  with the time  $\Pi_n$  spends in the state j, provided this state is visited. Given that the sequence of visited states is  $n = i_0 > i_1 > \cdots > i_{k-1} > i_k = 1$ , the total branch length  $L_n$  is distributed as  $\sum_{r=0}^{k-1} i_r T_{i_r}$  for  $n \in \mathbb{N} \setminus \{1\}$ .

For  $k \in \{1, \ldots, n\}$  and  $\hat{i} = (i_0, \ldots, i_k)$  with  $n = i_0 > i_1 > \cdots > i_{k-1} > i_k = 1$ , define the events  $A_{k,\hat{i}} := \{X_n = k, (\Pi_n(t_0), \ldots, \Pi_n(t_k)) = \hat{i}\}$ , where  $t_0 = 0$  and  $t_1 < t_2 < \cdots$  are the collision epochs. We have

$$\mathbb{E}(bL_n - X_n)^2 = \sum_{k,\hat{i}} \mathbb{P}\{A_{k,\hat{i}}\} \mathbb{E}\left(\sum_{r=0}^{k-1} (bi_r T_{i_r} - 1)\right)^2$$
$$= \sum_{k,\hat{i}} \mathbb{P}\{A_{k,\hat{i}}\} \left(\sum_{r=0}^{k-1} \mathbb{E}(bi_r T_{i_r} - 1)^2 + \sum_{r,s=0, r \neq s}^{k-1} \mathbb{E}(bi_r T_{i_r} - 1)(bi_s T_{i_s} - 1)\right).$$

Furthermore,  $\lambda_n = bn + O(\log n)$  as  $n \to \infty$  for a = 1 and b > 0 (see (A.5) below), which implies that  $|\mathbb{E}(bkT_k - 1)| = O(k^{-1}\log k)$  and  $\mathbb{E}(bkT_k - 1)^2 = 1 + O(k^{-1}\log k)$ . Therefore,

$$\mathbb{E}(bL_n - X_n)^2 \le \sum_{k,\hat{i}} \mathbb{P}\{A_{k,\hat{i}}\} \left( \sum_{r=2}^n \mathbb{E}(brT_r - 1)^2 + \left( \sum_{r=2}^n |\mathbb{E}(brT_r - 1)| \right)^2 \right)$$
  
=  $\sum_{k,\hat{i}} \mathbb{P}\{A_{k,\hat{i}}\}(n + O(\log^4 n))$   
=  $n + O(\log^4 n),$ 

and the convergence in  $L_2$  follows. Corollary 3.2 follows from the fact that, given  $L_n$ , the distribution of  $M_n$  is Poisson with mean  $rL_n$ . See Corollary 6.2 of [8] for details.

### 5.3. Proofs of Theorem 3.2 and Corollary 3.3

Let us verify (3.6) by induction on  $j \in \mathbb{N}$ . From (3.2), it follows that  $a_1 := \mathbb{E}X_1 = 0$  and  $a_n := \mathbb{E}X_n = 1 + \sum_{m=2}^{n-1} p_{n,m}^{(1)} a_m$ ,  $n \ge 2$ . In the following we apply the method of sequential approximations to the sequence  $(a_n)_{n \in \mathbb{N}}$ . The sequence  $(b_n)_{n \in \mathbb{N}}$ , defined via  $b_1 := 0$  and  $b_n := a_n - n/\log n$  for  $n \ge 2$ , satisfies the recursion

$$b_n = a_n - \frac{n}{\log n} = 1 + \sum_{m=2}^{n-1} p_{n,m}^{(1)} \left(\frac{m}{\log m} + b_m\right) - \frac{n}{\log n} = q_n + \sum_{m=2}^{n-1} p_{n,m}^{(1)} b_m$$

for  $n \ge 2$ , where  $q_n := 1 - n/\log n + \sum_{m=2}^{n-1} p_{n,m}^{(1)} m/\log m$ ,  $n \ge 2$ . By Corollary A.1 (applied with  $\alpha = 1$  and p = 1),

$$q_n = 1 - \frac{n}{\log n} + \left(\frac{n}{\log n} - 1 + \frac{m_1}{\log n} + O\left(\frac{1}{\log^2 n}\right)\right) = \frac{m_1}{\log n} + O\left(\frac{1}{\log^2 n}\right),$$

where  $m_1 := c_{b,1,1} = 2 + \Psi(b)$ . The sequence  $(c_n)_{n \in \mathbb{N}}$ , defined via  $c_1 := 0$  and  $c_n := b_n - m_1 n / \log^2 n$  for  $n \ge 2$ , therefore satisfies the recursion

$$c_n = b_n - m_1 \frac{n}{\log^2 n} = q_n + \sum_{m=2}^{n-1} p_{n,m}^{(1)} \left( m_1 \frac{m}{\log^2 m} + c_m \right) - m_1 \frac{n}{\log^2 n} = q'_n + \sum_{m=2}^{n-1} p_{n,m}^{(1)} c_m$$

for  $n \ge 2$ , where  $q'_n := q_n - m_1 n / \log^2 n + m_1 \sum_{m=2}^{n-1} p_{n,m}^{(1)} m / \log^2 m$ ,  $n \ge 2$ . By Corollary A.1 (applied with  $\alpha = 1$  and p = 2),

$$q'_{n} = q_{n} - m_{1} \frac{n}{\log^{2} n} + m_{1} \left( \frac{n}{\log^{2} n} - \frac{1}{\log n} + O\left(\frac{1}{\log^{2} n}\right) \right) = O\left(\frac{1}{\log^{2} n}\right)$$

since  $q_n = m_1/\log n + O(1/\log^2 n)$ . By Lemma A.2 (applied with  $\alpha = 1$  and p = 3), it follows that  $c_n = O(n/\log^3 n)$ . Thus, (3.6) holds for j = 1. Assume now that  $j \ge 2$ . From  $\mathbb{E}X_{I_n}^j = \mathbb{E}(X_n - 1)^j = \sum_{i=0}^{j-1} {j \choose i} (-1)^{j-i} \mathbb{E}X_n^i + \mathbb{E}X_n^j$ , it follows that

$$a_{n,j} := \mathbb{E}X_n^j = \sum_{i=0}^{j-1} \binom{j}{i} (-1)^{j-1-i} \mathbb{E}X_n^i + \mathbb{E}X_{I_n}^j = q_{n,j} + \sum_{m=2}^{n-1} p_{n,m}^{(1)} a_{m,j}$$

for  $n \ge 2$ , where  $q_{n,j} := \sum_{i=0}^{j-1} {j \choose i} (-1)^{j-1-i} \mathbb{E} X_n^i$ ,  $n \ge 2$ . Since, by induction, for all i < j,

$$\mathbb{E}X_n^i = \frac{n^i}{\log^i n} \left(1 + \frac{m_i}{\log n} + O\left(\frac{1}{\log^2 n}\right)\right),$$

it follows that (the summand for i = j - 1 asymptotically dominates the others)

$$q_{n,j} = \frac{jn^{j-1}}{\log^{j-1}n} \left( 1 + \frac{m_{j-1}}{\log n} + O\left(\frac{1}{\log^2 n}\right) \right).$$

Now apply the method of sequential approximations to the sequence  $(a_{n,j})_{n \in \mathbb{N}}$ . The sequence  $(b_{n,j})_{n \in \mathbb{N}}$ , defined via  $b_{1,j} := 0$  and  $b_{n,j} := a_{n,j} - n^j / \log^j n$  for  $n \ge 2$ , satisfies the recursion  $b_{n,j} = q'_{n,j} + \sum_{m=2}^{n-1} p_{n,m}^{(1)} b_{m,j}, n \ge 2$ , where

$$q'_{n,j} := q_{n,j} - \frac{n^j}{\log^j n} + \sum_{m=2}^{n-1} \frac{p_{n,m}^{(1)} m^j}{\log^j m}, \qquad n \ge 2.$$

By Corollary A.1 (applied with  $\alpha = j$  and p = j),

$$\begin{aligned} q_{n,j}' &= j \frac{n^{j-1}}{\log^{j-1} n} + j m_{j-1} \frac{n^{j-1}}{\log^{j} n} + O\left(\frac{n^{j-1}}{\log^{j+1} n}\right) - \frac{n^{j}}{\log^{j} n} \\ &+ \frac{n^{j}}{\log^{j} n} - j \frac{n^{j-1}}{\log^{j-1} n} + \kappa_{j} \frac{n^{j-1}}{\log^{j} n} + O\left(\frac{n^{j-1}}{\log^{j+1} n}\right) \\ &= j m_{j} \frac{n^{j-1}}{\log^{j} n} + O\left(\frac{n^{j-1}}{\log^{j+1} n}\right), \end{aligned}$$

where  $\kappa_j := c_{b,j,j}$  and  $m_j := m_{j-1} + \kappa_j/j$ . The sequence  $(c_{n,j})_{n \in \mathbb{N}}$ , defined via  $c_{1,j} := 0$ and  $c_{n,j} := b_{n,j} - m_j n^j / \log^{j+1} n$  for  $n \ge 2$ , therefore satisfies the recursion  $c_{n,j} = q_{n,j}'' + \sum_{m=2}^{n-1} p_{n,m}^{(1)} c_{m,j}, n \ge 2$ , where  $q_{n,j}'' := q_{n,j}' - m_j n^j / \log^{j+1} n + m_j \sum_{m=2}^{n-1} p_{n,m}^{(1)} m^j / \log^{j+1} m$ ,  $n \ge 2$ . By Corollary A.1 (applied with  $\alpha = j$  and p = j + 1),

$$q_{n,j}'' = jm_j \frac{n^{j-1}}{\log^j n} + O\left(\frac{n^{j-1}}{\log^{j+1} n}\right) - m_j \frac{n^j}{\log^{j+1} n} + m_j \left(\frac{n^j}{\log^{j+1} n} - j\frac{n^{j-1}}{\log^j n} + O\left(\frac{n^{j-1}}{\log^{j+1} n}\right)\right) = O\left(\frac{n^{j-1}}{\log^{j+1} n}\right).$$

By Lemma A.2 (applied with  $\alpha := j$  and with p := j+2), it follows that  $c_{n,j} = O(n^j/\log^{j+2} n)$ . Thus, (3.6) holds for j, which completes the induction and the proof of Theorem 3.2.

We now turn to the proof of Corollary 3.3. Let us first verify that the sequence  $(m_j)_{j \in \mathbb{N}_0}$ , recursively defined in Theorem 3.2, satisfies the inversion formula

$$\sum_{i=0}^{j} {j \choose i} (-1)^{j-i} m_i = \frac{(-1)^j}{j} \mathbf{B}(b, j-1), \qquad j \in \mathbb{N} \setminus \{1\}.$$
(5.4)

Using the formula  $\Psi(x+1) = \Psi(x) + 1/x$ ,  $x \in (0, \infty)$ , it is readily checked that  $\kappa_{j+1} - \kappa_j = 2 + \Psi(b+j)$ ,  $j \in \mathbb{N}_0$ . For all  $j \in \mathbb{N}_0$ , it follows that

$$\kappa_j = \sum_{i=0}^{j-1} (\kappa_{i+1} - \kappa_i) = \sum_{i=0}^{j-1} (2 + \Psi(b+i)) = 2j + \sum_{i=0}^{j-1} \Psi(b+i)$$

and

$$m_{j} = \sum_{l=1}^{j} (m_{l} - m_{l-1}) = \sum_{l=1}^{j} \frac{\kappa_{l}}{l} = \sum_{l=1}^{j} \left( 2 + \frac{1}{l} \sum_{i=0}^{l-1} \Psi(b+i) \right) = 2j + \sum_{i=0}^{j-1} \Psi(b+i) \sum_{l=i+1}^{j} \frac{1}{l}.$$
(5.5)

By (5.5), for  $j \in \{2, 3, ...\}$ ,

$$\begin{split} \sum_{i=0}^{j} {j \choose i} (-1)^{j-i} m_i &= \sum_{i=1}^{j} {j \choose i} (-1)^{j-i} \left( 2i + \sum_{k=0}^{i-1} \Psi(b+k) \sum_{l=k+1}^{i} \frac{1}{l} \right) \\ &= \sum_{i=1}^{j} {j \choose i} (-1)^{j-i} \sum_{k=0}^{i-1} \Psi(b+k) \sum_{l=k+1}^{i} \frac{1}{l} \\ &= \sum_{k=0}^{j-1} \Psi(b+k) \sum_{l=k+1}^{j} \frac{1}{l} \sum_{i=l}^{j} {j \choose i} (-1)^{j-i} \\ &= \sum_{k=0}^{j-1} \Psi(b+k) \sum_{l=k+1}^{j} \frac{1}{l} {j \choose l-1} (-1)^{j-l} \\ &= \frac{1}{j} \sum_{k=0}^{j-1} \Psi(b+k) \sum_{l=k+1}^{j} {j \choose l} (-1)^{j-l-k} . \end{split}$$

Substituting  $\binom{j-1}{k} = \binom{j-2}{k-1} + \binom{j-2}{k}$  and reordering with respect to  $\binom{j-2}{k}$  leads to

$$\begin{split} \sum_{i=0}^{j} \binom{j}{i} (-1)^{j-i} m_i &= \frac{1}{j} \sum_{k=0}^{j-2} (-1)^{j-2-k} \binom{j-2}{k} (\Psi(b+k+1) - \Psi(b+k)) \\ &= \frac{(-1)^j}{j} \sum_{k=0}^{j-2} (-1)^k \binom{j-2}{k} \frac{1}{b+k} \\ &= \frac{(-1)^j}{j} \mathbf{B}(b, j-1), \end{split}$$

where the last equality holds since

$$\sum_{k=0}^{n} \frac{(-1)^{k}}{b+k} \binom{n}{k} = \mathbf{B}(b, n+1)$$

for all  $n \in \mathbb{N}_0$ , which is, for example, readily verified by induction on  $n \in \mathbb{N}_0$ . Thus, (5.4) is established.

Thanks to Theorem 3.2 and the inversion formula (5.4), the proof of Corollary 3.3 is now straightforward. Basically, the same argument has been used by, e.g. Panholzer [28, p. 277]. Substituting the expansion for the ordinary moments given in (3.6) shows that

$$\begin{split} \mathbb{E}(X_n - \mathbb{E}X_n)^j \\ &= \sum_{i=0}^j \binom{j}{i} (-1)^{j-i} \mathbb{E}X_n^i (\mathbb{E}X_n)^{j-i} \\ &= \sum_{i=0}^j \binom{j}{i} (-1)^{j-i} \frac{n^i}{\log^i n} \left( 1 + \frac{m_i}{\log n} + O\left(\frac{1}{\log^2 n}\right) \right) \\ &\quad \times \left( \frac{n}{\log n} \left( 1 + \frac{m_1}{\log n} + O\left(\frac{1}{\log^2 n}\right) \right) \right)^{j-i} \\ &= \frac{n^j}{\log^j n} \sum_{i=0}^j \binom{j}{i} (-1)^{j-i} \left( 1 + \frac{m_i}{\log n} + O\left(\frac{1}{\log^2 n}\right) \right) \\ &\quad \times \left( 1 + \frac{(j-i)m_1}{\log n} + O\left(\frac{1}{\log^2 n}\right) \right) \\ &= \frac{n^j}{\log^j n} \sum_{i=0}^j \binom{j}{i} (-1)^{j-i} \left( 1 + \frac{(j-i)m_1 + m_i}{\log n} + O\left(\frac{1}{\log^2 n}\right) \right) \\ &= \frac{n^j}{\log^j n} \sum_{i=0}^j \binom{j}{i} (-1)^{j-i} + \frac{n^j}{\log^{j+1} n} \sum_{i=0}^j \binom{j}{i} (-1)^{j-i} ((j-i)m_1 + m_i) \\ &\quad + O\left(\frac{n^j}{\log^{j+2} n}\right) \\ &= \frac{n^j}{\log^{j+1} n} \frac{(-1)^j}{j} \mathbf{B}(b, j-1) + O\left(\frac{n^j}{\log^{j+2} n}\right), \end{split}$$

since  $\sum_{i=0}^{j} {j \choose i} (-1)^{j-i} = 0$ ,  $\sum_{i=0}^{j} {j \choose i} (-1)^{j-i} (j-i) = 0$ , and  $\sum_{i=0}^{j} {j \choose i} (-1)^{j-i} m_i = (-1)^j / j B(b, j-1)$  for  $j \ge 2$  by (5.4). This completes the proof of Corollary 3.3.

#### Appendix A

For each  $n \in \mathbb{N}$ , let  $(p_{n,k})_{0 \le k \le n}$  be an arbitrary probability distribution with  $p_{n,n} < 1$ . Define a sequence  $(a_n)_{n \in \mathbb{N}}$  as a (unique) solution to the recursion

$$a_n = r_n + \sum_{k=0}^n p_{n,k} a_k, \qquad n \in \mathbb{N},$$
(A.1)

with given  $r_n \ge 0$  and given initial value  $a_0 = a \ge 0$ .

**Lemma A.1.** ([14, Lemma 6.1].) Suppose that there exists a sequence  $(\psi_n)_{n \in \mathbb{N}}$  such that

- (C1)  $\liminf_{n\to\infty} \psi_n \sum_{k=0}^n (1-k/n) p_{n,k} > 0$ , and
- (C2) the sequence  $(r_k \psi_k / k)_{k \in \mathbb{N}}$  is nonincreasing.

Then  $a_n$ , defined by (A.1), satisfies

$$a_n = O\left(\sum_{k=1}^n \frac{r_k \psi_k}{k}\right) \quad as \ n \to \infty.$$

**Lemma A.2.** Let  $(a_n)_{n \in \mathbb{N}}$  be a sequence of real numbers satisfying the recursion  $a_1 = 0$ and  $a_n = q_n + \sum_{m=2}^{n-1} p_{n,m}^{(1)} a_m$ ,  $n \in \mathbb{N} \setminus \{1\}$ , for some given sequence  $(q_n)_{n \in \mathbb{N} \setminus \{1\}}$ , where  $p_{n,m}^{(1)}$ is defined via (3.4). If  $q_n = O(n^{\alpha-1}/\log^{p-1} n)$  for some given constants  $\alpha \in (0, \infty)$  and  $p \in [0, \infty)$ , then  $a_n = O(n^{\alpha}/\log^p n)$ .

*Proof.* Fix some  $\delta$  such that  $0 < \delta < \alpha$ . Set  $a'_n := |a_n|/n^{\delta}$  and  $q'_n := |q_n|/n^{\delta}$ . Then  $q'_n \le Mn^{\alpha-1-\delta}/\log^{p-1}n =: r_n$  for some M > 0 and all  $n \ge 2$ . Furthermore,

$$a'_{n} \le q'_{n} + \sum_{m=2}^{n-1} p_{n,m}^{(1)} \frac{|a_{m}|}{n^{\delta}} \le q'_{n} + \sum_{m=2}^{n-1} p_{n,m}^{(1)} \frac{|a_{m}|}{m^{\delta}} = q'_{n} + \sum_{m=2}^{n-1} p_{n,m}^{(1)} a'_{m} \le r_{n} + \sum_{m=2}^{n-1} p_{n,m}^{(1)} a'_{m}.$$

Set  $\psi_n := n/\log n$ . Then both conditions (C1) and (C2) hold. Hence,

$$a'_{n} = O\left(\sum_{k=2}^{n} \frac{k^{\alpha-1-\delta}}{\log^{p} k}\right) = O\left(\frac{n^{\alpha-\delta}}{\log^{p} n}\right) \text{ and } |a_{n}| = n^{\delta}a'_{n} = O\left(\frac{n^{\alpha}}{\log^{p} n}\right).$$

**Lemma A.3.** For the first decrement  $I_n$  of the Markov chain  $(\Pi_n)$  associated with the beta (a, b)-coalescent  $(a \in (0, 1] \text{ and } b > 0)$  and a random variable  $\xi$  with distribution  $(p_k^{(a)})_{k \in \mathbb{N}}$ ,

$$\sum_{k=1}^{n-1} k^q |\mathbb{P}\{I_n = k\} - \mathbb{P}\{\xi = k\}| = O(n^{a+q-2}),$$
(A.2)

whenever  $0 < q \leq 1$  and q + a > 1.

*Proof.* For the beta (a, b)-coalescents, (1.1) reads

$$\lambda_{n,k+1} = \int_0^1 x^{k-1} (1-x)^{n-k-1} \Lambda(\mathrm{d}x) = \frac{\mathrm{B}(a+k-1,n-k+b-1)}{\mathrm{B}(a,b)}.$$

Using the known estimate  $|\Gamma(n+c)/\Gamma(n+d) - n^{c-d}| \le M_{c,d}n^{c-d-1}$ ,  $n \ge 2$  and c, d > -2, for the gamma function (see Formula (6.1.47) of [1]), we obtain

$$\binom{n}{k+1}\lambda_{n,k+1} = \binom{n}{k+1}\frac{B(a+k-1,n-k+b-1)}{B(a,b)}$$
(A.3)  
$$= \frac{\Gamma(n+1)\Gamma(a+k-1)\Gamma(n-k+b-1)}{\Gamma(k+2)\Gamma(n-k)\Gamma(n+a+b-2)B(a,b)}$$
$$= \frac{\Gamma(a+k-1)}{(k+1)! B(a,b)}(n^{3-a-b} + O(n^{2-a-b}))((n-k)^{b-1} + O((n-k)^{b-2})),$$

uniformly for  $1 \le k \le n - 1$  and  $n \ge 2$ .

Using (3.1) with  $\Lambda$  given by (1.2), we infer (see also Corollary 2 of [13]) that

$$\lambda_n = \frac{\Gamma(a)}{(2-a)\mathbf{B}(a,b)} n^{2-a} + O(n^{1-a}) = \frac{\Gamma(a)}{(2-a)\mathbf{B}(a,b)} n^{2-a} (1+O(n^{-1}))$$
(A.4)

when  $a \in (0, 1)$  and b > 0, and

$$\lambda_n = bn + O(\log n) \tag{A.5}$$

when a = 1 and b > 0. Hence, for  $0 < a < 1, b > 0, n \ge 2$ , and k = 1, ..., n - 1,

$$\begin{split} p_{n,n-k}^{(a)} &= \frac{(2-a)\Gamma(a+k-1)}{\Gamma(a)(k+1)!} n^{1-b}((n-k)^{b-1} + O((n-k)^{b-2})) \left(1 + O\left(\frac{1}{n}\right)\right) \\ &= p_k^{(a)} \left( \left(1 - \frac{k}{n}\right)^{b-1} + O\left(\frac{1}{n}\left(1 - \frac{k}{n}\right)^{b-2}\right) \right) \left(1 + O\left(\frac{1}{n}\right)\right) \\ &= p_k^{(a)} \left( \left(1 - \frac{k}{n}\right)^{b-1} + O\left(\frac{1}{n}\left(1 - \frac{k}{n}\right)^{b-2}\right) \right) \\ &= p_k^{(a)} \left(1 - \frac{k}{n}\right)^{b-1} + O\left(p_k^{(a)}\frac{1}{n}\left(1 - \frac{k}{n}\right)^{b-2}\right). \end{split}$$

Analogously, for a = 1,

$$p_{n,n-k}^{(1)} = p_k^{(1)} \left( \left( 1 - \frac{k}{n} \right)^{b-1} + O\left( \frac{1}{n} \left( 1 - \frac{k}{n} \right)^{b-2} \right) \right) (1 + O(n^{-1} \log n))$$

$$= p_k^{(1)} \left( \left( 1 - \frac{k}{n} \right)^{b-1} + O\left( \frac{1}{n} \left( 1 - \frac{k}{n} \right)^{b-2} \right) + O\left( \frac{1}{n} \log n \left( 1 - \frac{k}{n} \right)^{b-1} \right) \right)$$

$$= p_k^{(1)} \left( 1 - \frac{k}{n} \right)^{b-1} + O\left( p_k^{(1)} \frac{1}{n} \left( 1 - \frac{k}{n} \right)^{b-2} \right) + O\left( p_k^{(1)} \frac{1}{n} \log n \left( 1 - \frac{k}{n} \right)^{b-1} \right).$$

Substituting these expansions into the left-hand side of (A.2) gives

$$\sum_{k=1}^{n-1} k^{q} |\mathbb{P}\{I_{n} = k\} - \mathbb{P}\{\xi = k\}| \leq \sum_{k=1}^{n-1} p_{k}^{(a)} k^{q} \left| \left(1 - \frac{k}{n}\right)^{b-1} - 1 \right| + \frac{c_{1}}{n} \sum_{k=1}^{n-1} p_{k}^{(a)} k^{q} \left(1 - \frac{k}{n}\right)^{b-2}$$
$$=: S_{1}(a, n) + S_{2}(a, n) \quad \text{for } 0 < a < 1,$$

and

$$\sum_{k=1}^{n-1} k^{q} |\mathbb{P}\{I_{n} = k\} - \mathbb{P}\{\xi = k\}| \le S_{1}(1, n) + S_{2}(1, n) + \frac{c_{2} \log n}{n} \sum_{k=1}^{n-1} p_{k}^{(1)} k^{q} \left(1 - \frac{k}{n}\right)^{b-1}$$
$$=: S_{1}(1, n) + S_{2}(1, n) + S_{3}(1, n) \quad \text{for } a = 1.$$

Here and hereafter  $c_1, c_2, \ldots$  denote some positive constants whose values are of no importance. Our aim is to show that  $S_i(a, n) = O(n^{q+a-2})$  for i = 1, 2 and  $S_3(1, n) = O(n^{q-1})$ . By virtue of  $p_k^{(a)} \le c_3 k^{a-3}$  for all  $k \in \mathbb{N}$ , we infer that

$$S_{1}(a,n) \leq c_{3} \sum_{k=1}^{n-1} k^{a+q-3} \left| \left( 1 - \frac{k}{n} \right)^{b-1} - 1 \right|$$
  
$$= c_{3} \sum_{k=1}^{[n/2]} k^{a+q-3} \left| \left( 1 - \frac{k}{n} \right)^{b-1} - 1 \right| + c_{3} \sum_{k=[n/2]+1}^{n-1} k^{a+q-3} \left| \left( 1 - \frac{k}{n} \right)^{b-1} - 1 \right|$$
  
$$\leq \frac{c_{4}}{n} \sum_{k=1}^{[n/2]} k^{a+q-2} + c_{3} n^{a+q-2} \left( \frac{1}{n} \sum_{k=[n/2]+1}^{n-1} \left( \frac{k}{n} \right)^{a+q-3} \left| \left( 1 - \frac{k}{n} \right)^{b-1} - 1 \right| \right),$$

where the inequality  $|(1-x)^{b-1}-1| \le c_5 x$ ,  $x \in [0, \frac{1}{2}]$ , has been utilized. The expression in the parentheses converges to  $\int_{1/2}^{1} x^{a+q-3} |(1-x)^{b-1}-1| dx < \infty$ . Hence,  $S_1(a, n) = O(n^{q+a-2})$ . Similarly,

$$\begin{split} S_{2}(a,n) &\leq \frac{c_{6}}{n} \sum_{k=1}^{n-1} k^{a+q-3} \left(1 - \frac{k}{n}\right)^{b-2} \\ &= \frac{c_{6}}{n} \sum_{k=1}^{[n/2]} k^{a+q-3} \left(1 - \frac{k}{n}\right)^{b-2} + \frac{c_{6}}{n} \sum_{k=[n/2]+1}^{n-1} k^{a+q-3} \left(1 - \frac{k}{n}\right)^{b-2} \\ &\leq \frac{c_{6}}{n} \sum_{k=1}^{[n/2]} k^{a+q-3} \left(1 - \frac{k}{n}\right)^{b-2} + c_{6} \sum_{k=[n/2]+1}^{n-1} k^{a+q-3} \left(1 - \frac{k}{n}\right)^{b-1} \\ &= \frac{c_{6}}{n} \sum_{k=1}^{[n/2]} k^{a+q-3} \left(1 - \frac{k}{n}\right)^{b-2} + c_{6} n^{a+q-2} \left(\frac{1}{n} \sum_{k=[n/2]+1}^{n-1} \left(\frac{k}{n}\right)^{a+q-3} \left(1 - \frac{k}{n}\right)^{b-1}\right) \\ &= O(n^{a+q-2}), \end{split}$$

since the first term is  $O(n^{-1})$  and the second term is  $O(n^{a+q-2})$  by the same reasoning as for  $S_1(a, n)$ . Finally,

$$S_{3}(1,n) \leq \frac{c_{7}\log n}{n} \sum_{k=1}^{n-1} k^{q-2} \left(1 - \frac{k}{n}\right)^{b-1}$$
$$\leq \frac{c_{7}\log n}{n} \sum_{k=1}^{n-1} k^{q-2} \left| \left(1 - \frac{k}{n}\right)^{b-1} - 1 \right| + \frac{c_{7}\log n}{n} \sum_{k=1}^{n-1} k^{q-2}$$
$$= O(n^{q-2}\log n) + O(n^{-1}\log n),$$

in view of the estimate for  $S_1(a, n)$ . Thus,  $S_3(1, n) = O(n^{q-1})$ . This completes the proof.

We provide a basic lemma concerning the total rates of the beta (1, b)-coalescent.

**Lemma A.4.** The total rates  $\lambda_n$ ,  $n \in \mathbb{N}$ , of the beta (1, b)-coalescent are given by

$$\lambda_n = b \sum_{k=1}^{n-1} \frac{k}{b+k-1} = b(n-1) - b(b-1)(\Psi(n+b-1) - \Psi(b)), \qquad n \in \mathbb{N}.$$
 (A.6)

Moreover, the total rates have the asymptotic expansion

$$\lambda_n = bn - b(b-1)\log n - b + b(b-1)\Psi(b) + O(n^{-1}) \quad as \ n \to \infty,$$
(A.7)

and the inverse of the total rate  $\lambda_n$  has the asymptotic expansion

$$\frac{1}{\lambda_n} = \frac{1}{bn} \left( 1 + (b-1)\frac{\log n}{n} + \frac{1 - (b-1)\Psi(b)}{n} + O\left(\frac{\log^2 n}{n^2}\right) \right) \quad as \ n \to \infty.$$
(A.8)

*Proof.* Equation (A.6) is known (see, for example, [19, Equation (19)]). Expansion (A.7) follows from (A.6), since  $\Psi(n+b-1) = \log n + O(n^{-1})$  as  $n \to \infty$ . Equation (A.8) follows from

$$\frac{bn}{\lambda_n} - 1 - (b-1)\frac{\log n}{n} - \frac{1 - (b-1)\Psi(b)}{n}$$
$$= \frac{bn^2 - \lambda_n(n+(b-1)\log n + 1 - (b-1)\Psi(b))}{n\lambda_n}$$
$$= \frac{O(\log^2 n)}{n\lambda_n}$$
$$= O\left(\frac{\log^2 n}{n^2}\right),$$

where the last equality holds since  $\lambda_n \sim bn$ , and the equality before follows by substituting the  $\lambda_n$  term in the numerator for the expression given in (A.7) and multiplying everything out.

The next lemma provides asymptotic expansions as  $n \to \infty$  for the sum

$$s_n(p,\alpha) := \sum_{m=2}^{n-1} \frac{m^{\alpha}}{(n-m)(n-m+1)\log^p m}, \qquad p \in [0,\infty), \, \alpha \in \mathbb{R}.$$
 (A.9)

**Lemma A.5.** Fix  $p \in [0, \infty)$ . Then, as  $n \to \infty$ , the sum  $s_n(p, \alpha)$  defined in (A.9) satisfies  $s_n(p, \alpha) = O(n^{\alpha}/\log^p n)$  for  $\alpha \in (-2, \infty)$ ,

$$s_n(p,\alpha) = \frac{n^{\alpha}}{\log^p n} \left( 1 + O\left(\frac{\log n}{n}\right) \right) \text{ for } \alpha \in (-1,\infty),$$

and

$$s_n(p,\alpha) = \frac{n^{\alpha}}{\log^p n} \left( 1 - \alpha \frac{\log n}{n} + \frac{\alpha \Psi(\alpha) + p}{n} + O\left(\frac{1}{n \log n}\right) \right) \quad for \ \alpha \in (0,\infty).$$

For a proof of Lemma A.5, we refer the reader to [16], which is a preprint version of this article. The following corollary provides an asymptotic expansion which is a key tool in the proof of Theorem 3.2.

**Corollary A.1.** Fix  $\alpha \in [1, \infty)$  and  $p \in [0, \infty)$ . For the beta (1, b)-coalescent with parameter  $b \in (0, \infty)$ ,

$$\sum_{m=2}^{n-1} p_{n,m}^{(1)} \frac{m^{\alpha}}{\log^p m} = \frac{n^{\alpha}}{\log^p n} \left( 1 - \alpha \frac{\log n}{n} + \frac{c_{b,\alpha,p}}{n} + O\left(\frac{1}{n\log n}\right) \right) \quad as \ n \to \infty, \quad (A.10)$$

where  $c_{b,\alpha,p} := (\alpha + b - 1)\Psi(\alpha + b - 1) + p + 1 + (1 - b)\Psi(b) = (\alpha + b - 1)\Psi(\alpha + b) + p - (b - 1)\Psi(b).$ 

**Remark A.1.** The following proof shows that Corollary A.1 holds even for the slightly larger range of parameters  $\alpha$ ,  $b \in (0, \infty)$  satisfying  $\alpha + b - 1 > 0$ . However, we need Corollary A.1 only for  $\alpha \in [1, \infty)$  and  $b \in (0, \infty)$ , in which case  $\alpha + b - 1 > 0$  automatically holds.

*Proof of Corollary A.1.* Let  $g_{nm} := \lambda_n \mathbb{P}\{I_n = n - m\}$  denote the rate at which the block counting process moves from state n to state  $m \in \{1, ..., n - 1\}$ . It suffices to verify that

$$\sum_{m=2}^{n-1} g_{nm} \frac{m^{\alpha}}{\log^{p} m} = \frac{bn^{\alpha+1}}{\log^{p} n} \left( 1 - (\alpha + b - 1) \frac{\log n}{n} + \frac{(\alpha + b - 1)\Psi(\alpha + b - 1) + p}{n} + O\left(\frac{1}{n \log n}\right) \right),$$
(A.11)

since (A.10) then follows from  $p_{n,m}^{(1)} = g_{nm}/\lambda_n$  by multiplying (A.11) with (A.8). Note that

$$g_{nm} = b \frac{n!}{\Gamma(b+n-1)} \frac{1}{(n-m)(n-m+1)} \frac{\Gamma(b+m-1)}{(m-1)!}, \qquad 1 \le m < n.$$

Since the first fraction has expansion

$$\frac{n!}{\Gamma(b+n-1)} = \frac{1}{n^{b-2}} \left( 1 - {\binom{b-1}{2}} \frac{1}{n} + O\left(\frac{1}{n^2}\right) \right), \tag{A.12}$$

it hence suffices to verify that

$$\sum_{m=2}^{n-1} \frac{1}{(n-m)(n-m+1)} \frac{\Gamma(b+m-1)}{(m-1)!} \frac{m^{\alpha}}{\log^{p} m}$$
$$= \frac{n^{\alpha+b-1}}{\log^{p} n} \left(1 - (\alpha+b-1)\frac{\log n}{n} + \frac{\binom{b-1}{2} + (\alpha+b-1)\Psi(\alpha+b-1) + p}{n} + O\left(\frac{1}{n\log n}\right)\right),$$
(A.13)

since (A.11) then follows by multiplying (A.13) with (A.12). Thus, it remains to verify (A.13). Since, for all  $m \in \mathbb{N}$  and all  $b \in (0, \infty)$ , the Pochhammer-like expression  $\Gamma(b+m-1)/(m-1)!$  appearing on the left-hand side of (A.13) is bounded below and above by

$$m^{b-1} + {\binom{b-1}{2}}m^{b-2} \le \frac{\Gamma(b+m-1)}{(m-1)!} \le m^{b-1} + {\binom{b-1}{2}}m^{b-2} + K_b m^{b-3},$$

where  $K_b := \Gamma(b) - 1 - {\binom{b-1}{2}}$ , (A.13) follows by substituting these lower and upper bounds into the left-hand side of (A.13), then applying the last expansion in Lemma A.5 with  $\alpha$  replaced

by  $\alpha + b - 1 > 0$ , and noting that

$$\sum_{m=2}^{n-1} \frac{m^{\alpha+b-2}}{(n-m)(n-m+1)\log^p m} = \frac{n^{\alpha+b-2}}{\log^p n} \left(1 + O\left(\frac{\log n}{n}\right)\right)$$

and

$$\sum_{m=2}^{n-1} \frac{m^{\alpha+b-3}}{(n-m)(n-m+1)\log^p m} = O\left(\frac{n^{\alpha+b-3}}{\log^p n}\right),$$

again by Lemma A.5. This completes the proof.

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