Precise asymptotics of longest cycles in random permutations without macroscopic cycles

VOLKER BETZ1,* JULIAN MÜHLBAUER1 HELGE SCHÄFER1,** and DIRK ZEINDLER2

1Fachbereich Mathematik, Arbeitsgruppe Stochastik, Schloßgartenstrasse 7, 64289 Darmstadt
E-mail: *betz@mathematik.tu-darmstadt.de; **hschaefer@mathematik.tu-darmstadt.de
2Lancaster University, Mathematics and Statistics, Fylde College, Bailrigg, Lancaster LA1 4YF, UK E-mail: d.zeindler@lancaster.ac.uk,

We consider Ewens random permutations of length \( n \) conditioned to have no cycle longer than \( n^\beta \) with \( 0 < \beta < 1 \) and study the asymptotic behaviour as \( n \to \infty \). We obtain very precise information on the joint distribution of the lengths of the longest cycles; in particular we prove a functional limit theorem where the cumulative number of long cycles converges to a Poisson process in the suitable scaling. Furthermore, we prove convergence of the total variation distance between joint cycle counts and suitable independent Poisson random variables up to a significantly larger maximal cycle length than previously known. Finally, we remove a superfluous assumption from a central limit theorem for the total number of cycles proved in an earlier paper.

Keywords: random permutations, Ewens measure, long cycles, functional limit theorem, Total variation distance, cycle structure.

1. Introduction

In uniform random permutations, long cycles occupy almost all the available space. Indeed, it is a standard textbook exercise to show that in a permutation of length \( n \), the probability to find an index \( i \) in a cycle of length \( k \) is equal to \( 1/n \), which in turn means that cycles of a length below volume order play no role asymptotically as \( n \to \infty \). Of course, much more is known about uniform (and Ewens) random permutations, including the precise distribution of long and short cycles. We refer to [1] and the references therein.

It is interesting to see how the behaviour of random permutations changes when the uniform measure is changed in a way that favours short cycles. Various such models have been studied in recent years. Many of them are motivated by the model of spatial random permutations [5], which by its close connections to Bose-Einstein condensation [25] has a significant physical relevance. In this model, a spatial structure is superimposed on the permutations, and the importance of that spatial structure is measured by an order parameter which physically is the temperature. It is conjectured that this order parameter mediates a phase transition between a regime of only short cycles and a
regime of coexistence of long and short cycles. Despite some successes in the explicitly solvable annealed case without interaction between different cycles [7], and significant recent progress (using the method of reflection positivity) in a closely related model with such interaction [18, 24], many of the most relevant questions in spatial random permutations remain to be answered.

A somewhat more direct and in general easier to analyse way to suppress long cycles is to introduce cycle weights or hard constraints on cycle numbers. Cycle weights appear in an (uncontrolled) approximation of the interacting Bose gas by a variant of the free one [6], but have also been studied intensively in their own right, both in cases where the cycle weights do not depend on the system size $n$ [8, 13], and in cases where they do [10, 12]. In the latter case, it has been shown in the cited papers that one recovers the model treated in [7] by a suitable choice of cycle weights, and the methods of analytic combinatorics used in [10, 12] yield very precise information about the asymptotic cycle distribution in various regimes.

The present paper deals with the other option of constraining permutations, namely to completely disallow certain cycle lengths. Again, a distinction has to be made between cases where the set of disallowed cycle lengths is independent of the permutation length $n$, and those where it depends on $n$. In the first case, a significant amount of information has been obtained in the works of Yakymiv (see e.g. [26, 27]); our interest lies in the second case. Using precise asymptotic results by Manstavičius and Petuchovas [20], in [3, 4] we investigated the case where a permutation of length $n$ is prevented from having any cycles above a threshold $\alpha(n)$ that grows strictly slower than volume order. While the results in these papers were reasonably detailed, some interesting questions and fine details have been left out.

It is the purpose of the present paper to settle a significant portion of them. We will describe our results in detail in the next section. Here, we only briefly sketch what is new.

One difference to [3] is that we generalise the base model we constrain, from uniform random permutations to the model of Ewens permutations. The latter originally appeared in population genetics, see [14], but has now become a rather standard model of random permutations. It shares many features and techniques with uniform permutations, and classical results about uniform and Ewens random permutations include convergence of joint cycle counts towards independent Poisson random variables in total variation distance [2], the convergence of the renormalized cycle structure towards a Poisson-Dirichlet distribution [17, 23], and a central limit theorem for cumulative cycle counts [11].

In the context of the methods we use, the difference between the Ewens measure and uniform random permutations is not large, see [22] for details. What should be considered the main contribution of the present paper compared to [3, 4] are the following three items: firstly, we obtain much more precise asymptotics for the distribution of the longest cycles in various regimes (Propositions 2.2 and 2.3, and Theorem 2.5). For this, we consider the number of cycles of length $\alpha(n)$ and distinguish the cases where this number is diverging, is converging to a positive number and is vanishing. Secondly, we extend the validity of the joint Poisson approximation (in variation distance) to the
whole regime of cycles of length \( o(\alpha(n)) \) (Theorem 2.8). We achieve this by replacing the Poisson random variables with fixed expectation by Poisson random variables with an expectation depending on \( n \). Finally, we remove a spurious additional assumption for the central limit theorem for cycle numbers that was present in [4], see Theorem 2.9. We achieve this by a more precise saddle point solution.

The paper is organised as follows: in Section 2, we introduce the model, give our results and compare them to previously existing ones. In Section 3, we prove those results.

2. Model and Results

2.1. The symmetric group and the Ewens measure

For \( n \in \mathbb{N} \), let \( S_n \) be the group of all permutations of the set \( \{1, \ldots, n\} \). For \( \sigma \in S_n \) and \( m \in \mathbb{N} \), we denote by \( C_m(\sigma) \) the number of cycles of length \( m \) in the cycle decomposition of \( \sigma \) into disjoint cycles. Note that we typically write \( C_m \) instead of \( C_m(\sigma) \). Let \( n \mapsto \alpha(n) \) satisfy the condition

\[
  n^{a_1} \leq \alpha(n) \leq n^{a_2}
\]

with \( a_1, a_2 \in (0,1) \). We denote by \( S_{n,\alpha} \) the subset of \( S_n \) of all permutations \( \sigma \) for which all cycles in the cycle decomposition of \( \sigma \) have length at most \( \alpha(n) \). In other words, \( \sigma \in S_{n,\alpha} \) if and only if \( C_m(\sigma) = 0 \) for \( m > \alpha(n) \). For \( \vartheta > 0 \), the Ewens measure on \( S_n \) with parameter \( \vartheta \) is defined as

\[
  P_n[\sigma] := \frac{\prod_{m=1}^{n} \vartheta^{C_m(\sigma)}}{\vartheta(\vartheta + 1) \cdots (\vartheta + n - 1)}. \tag{2.2}
\]

Note that the case \( \vartheta = 1 \) corresponds to the uniform measure. Further, let \( P_{n,\alpha} \) denote the measure on \( S_{n,\alpha} \) obtained by conditioning \( P_n \) on \( S_{n,\alpha} \), i.e.

\[
  P_{n,\alpha}[A] := P_n[A | S_{n,\alpha}] \quad \text{for all } A \subset S_{n,\alpha}. \tag{2.3}
\]

Inserting the definition \( P_n \), we obtain for \( \sigma \in S_{n,\alpha} \) that

\[
  P_{n,\alpha}[\sigma] = \frac{\prod_{m=1}^{n} \vartheta^{C_m(\sigma)}}{Z_{n,\alpha} n!} \quad \text{with} \quad Z_{n,\alpha} = \frac{1}{n!} \sum_{\sigma \in S_{n,\alpha}} \prod_{m=1}^{n} \vartheta^{C_m(\sigma)}. \tag{2.4}
\]

Also, we write \( \mathbb{E}_n \) for the expectation with respect to \( P_n \) and \( \mathbb{E}_{n,\alpha} \) for the expectation with respect to \( P_{n,\alpha} \).

2.2. Notation

If two sequences \( (a_n) \) and \( (b_n) \) are asymptotically equivalent, i.e. if \( \lim_{n \to \infty} a_n/b_n = 1 \), we write \( a_n \sim b_n \). Further, we write \( a_n \approx b_n \) when there exist constants \( c_1, c_2 > 0 \) such that

\[
  c_1 b_n \leq a_n \leq c_2 b_n \tag{2.5}
\]
for large $n$. We also use the usual $O$ and $o$ notation, i.e. $f(n) = O(g(n))$ means that there exists some constant $c > 0$ so that $|f(n)| \leq c|g(n)|$ for large $n$, while $f(n) = o(g(n))$ means that for all $c > 0$ there exists $n_c \in \mathbb{N}$ so that the inequality $|f(n)| \leq c|g(n)|$ holds for all $n > n_c$. We further say that

$$f_n(t) = O(g_n(t))$$

uniformly in $t \in T_n$ as $n \to \infty$

if there are constants $c, N > 0$ such that $\sup_{t \in T_n} |f_n(t)| \leq c|g_n(t)|$ for all $n \geq N$.

### 2.3. Expected cycle counts

Here we recall some of the results from [4] and [22] that are crucial for the following. Let $x_{n,\alpha}$ be the unique positive solution of the equation

$$n = \vartheta \sum_{j=1}^{\alpha(n)} x_{n,\alpha}^j,$$  \hspace{1cm} (2.6)

and

$$\mu_m(n) := \vartheta \frac{x_{n,\alpha}^m}{m}.$$ \hspace{1cm} (2.7)

For the case where $m$ is replaced by an integer-valued sequence $(m(n))_{n \in \mathbb{N}}$, we simplify notation and write $\mu_m(n)$ instead of $\mu_{m(n)}(n)$. For any such sequence that satisfies $m(n) \leq \alpha(n)$, we have

$$\mathbb{E}_{n,\alpha} [C_m(n)] \sim \mu_m(n) \quad \text{as} \quad n \to \infty.$$ \hspace{1cm} (2.8)

This was proven for $\vartheta = 1$ in [4, Proposition 2.1], and for $\vartheta \neq 1$ in [22] along the same lines. In view of (2.8) it is clear that we are interested in information about the asymptotics of solutions to equations like (2.6). The following result provides it:

**Lemma 2.1.** Let $0 < c_1 < c_2 < \infty$ be fixed, but arbitrary real numbers. For $c \in [c_1, c_2]$, let $x_{n,\alpha}(c)$ be the solution of

$$cn = \vartheta \sum_{j=1}^{\alpha(n)} (x_{n,\alpha}(c))^j.$$ \hspace{1cm} (2.9)

We then have uniformly in $c \in [c_1, c_2]$ as $n \to \infty$

$$\alpha(n) \log(x_{n,\alpha}(c)) = \log \left( \frac{cn}{\vartheta \alpha(n)} \log \left( \frac{cn}{\vartheta \alpha(n)} \right) \right) + O \left( \frac{\log \log(n)}{\log(n)} \right).$$ \hspace{1cm} (2.10)

In particular,

$$x_{n,\alpha}(c) \geq 1, \quad \lim_{n \to \infty} x_{n,\alpha}(c) = 1 \quad \text{and} \quad (x_{n,\alpha}(c))^\alpha(n) \sim \frac{cn}{\vartheta \alpha(n)} \log \left( \frac{cn}{\vartheta \alpha(n)} \right).$$
for large $n$. Furthermore,

\[
\sum_{j=1}^{\alpha(n)} j (x_{n,\alpha(c)})^j \sim \frac{cn}{\vartheta} \alpha(n). \tag{2.11}
\]

Lemma 2.1 is a special case of [20, Lemma 9] and follows immediately by inserting our assumptions in [20, Lemma 9]. We thus omit the proof.

### 2.4. Asymptotics of longest cycles

The first set of results that we present deals with the asymptotic (joint) distribution of the longest cycles under the measure $P_{n,\alpha}$. Let $\ell_k = \ell_k(\sigma)$ denote the length of the $k$-th longest cycle of the permutation $\sigma$. We already know that for fixed $K \in \mathbb{N}$, under the probability measures $P_{n,\alpha}$, we have as $n \to \infty$

\[
\frac{1}{\alpha(n)} (\ell_1, \ell_2, \ldots, \ell_K) \xrightarrow{d} (1, 1, \ldots, 1), \tag{2.12}
\]

where $\xrightarrow{d}$ denotes convergence in distribution (see equation (2.14) in [4] or [22]). We will significantly improve on this information.

It turns out that the behaviour of the longest cycles depends on the expected length given in (2.8). In other words, we have to look at the behaviour of $\mu_{\alpha(n)}$ in the three regimes $\mu_{\alpha(n)} \to \infty$, $\mu_{\alpha(n)} \to \mu$ with $\mu > 0$, and $\mu_{\alpha(n)} \to 0$.

A discussion about which regime happens when in case of $\alpha(n) = n^{\beta}$ can be found in Section 2.2 of [4].

We start with the simplest case $\mu_{\alpha(n)} \to \infty$. This case only occurs if $\alpha(n) = o(n \log n)^{\frac{1}{2}}$, see Proposition 2.4 below. In this case, the distribution of the random vector $(\ell_1, \ldots, \ell_K)$ becomes degenerate:

**Proposition 2.2.** Suppose that $\mu_{\alpha(n)} \to \infty$. Then, for each $K \in \mathbb{N}$, we have

\[
\lim_{n \to \infty} P_{n,\alpha} \left[ (\ell_1, \ell_2, \ldots, \ell_K) \neq (\alpha(n), \alpha(n), \ldots, \alpha(n)) \right] = 0.
\]

A similar proposition was proven in [4, Theorem 2.8] and [22] under the additional assumption that $\alpha(n) \geq n^{\frac{1}{2} + \delta}$ for $\delta > 0$. The reason why we can omit this assumption here is our improved central limit theorem, Theorem 2.9. We give the proof of Proposition 2.2 in Section 3.2.

Next, we now look at the case $\mu_{m(n)} \to \mu$ with $\mu > 0$. We find
Proposition 2.3. Suppose that $\mu_{\alpha(n)} \to \mu$ with $\mu > 0$ as $n \to \infty$. We then have for all $d \in \mathbb{N}_0$ and all $k \in \mathbb{N}$ that
\[
\mathbb{P}_{n,\alpha}[\ell_k = \alpha(n) - d] \xrightarrow{n \to \infty} \frac{1}{\Gamma(k)} \int_{d\mu}^{(d+1)\mu} v^{k-1} e^{-v} dv.
\] (2.13)

In other words, $\alpha(n) - \ell_k$ converges in distribution to $\lfloor \mu^{-1} X \rfloor$, where $X$ is a gamma-distributed random variable with parameters $k$ and 1 and $\lfloor x \rfloor = \max\{n \in \mathbb{Z}; n \leq x\}$.

The proof of this proposition is given in Section 3.3. Moreover, the proof allows for deriving the joint distribution of the longest cycles, but the notation of results in this case is cumbersome.

Finally, we have the case where the expected number of cycles vanishes. Here we obtain the most interesting results, namely a functional convergence of the cumulative numbers of long cycles to a Poisson process, on the correct scale. By considering the jump times of this Poisson process, we establish limit theorems for $\ell_k$. Let us start with a small observation.

Proposition 2.4. We have, as $n \to \infty$,
\[
\mu_{\alpha(n)} \approx \frac{n \log n}{(\alpha(n))^2}.
\] (2.14)

Proof. Inserting the definition of $\mu_{\alpha(n)}$, see (2.7), and using Lemma 2.1, we obtain
\[
\mu_{\alpha(n)} = \frac{n \alpha(n)}{\alpha(n)} \sim \frac{n}{(\alpha(n))^2} \log \left( \frac{n}{\theta \alpha(n)} \right) \approx \frac{n \log n}{(\alpha(n))^2}.
\] (2.15)

This completes proof of this proposition. \hfill \qed

This proposition immediately implies that $\mu_{\alpha(n)} \to 0$ if and only if $\frac{n \log n}{(\alpha(n))^2} \to 0$ as $n \to \infty$. We now define
\[
d_t(n) := \max \left\{ \alpha(n) - \left\lfloor \frac{t}{\mu_{\alpha(n)}} \right\rfloor, 0 \right\}.
\] (2.16)

Note that $d_t(n) = \alpha(n)(1 + o(1))$ and $\left\lfloor \frac{t}{\mu_{\alpha(n)}} \right\rfloor \to \infty$ if $\mu_{\alpha(n)} \to 0$ for fixed $t$. We now have

Theorem 2.5. Suppose that $\mu_{\alpha(n)} \to 0$ and define for $t \geq 0$
\[
P_t := \sum_{j=d_t(n)+1}^{\alpha(n)} C_j.
\]

Then the stochastic process $\{P_t, t \geq 0\}$ converges under $\mathbb{P}_{n,\alpha}$ as $n \to \infty$ weakly in $\mathcal{D}[0,\infty)$ to a Poisson process with parameter 1, where $\mathcal{D}[0,\infty)$ denotes the space of càdlàg-functions.
This theorem is proved in Section 3.4. It immediately implies the following corollary.

**Corollary 2.6.** Let $K \in \mathbb{N}$ be given, $\alpha(n)$ be as in (2.1) and suppose that $\mu_{\alpha(n)} \to 0$. We have convergence in distribution of

$$\mu_{\alpha(n)} \cdot (\alpha(n) - \ell_1, \ell_2 - \ell_1, \ldots, \ell_{K-1})$$

under $\mathbb{P}_{n,\alpha}$ to independent exponentially distributed random variables with parameters 1. In particular, $\mu_{\alpha(n)}(\alpha(n) - \ell_k)$ converges in distribution to a gamma-distributed random variable with parameters $k$ and 1.

**Proof.** The claim is a consequence of the convergence established in the proof of Theorem 2.5 since the limit distribution is the distribution of the jump times of the Poisson process (see, e.g. [19, p. 5]). \qed

### 2.5. Total variation distance

Here we study the joint behaviour of the cycle counts $C_m$ in the region $m = o(\alpha(n))$. Recall that the total variation distance of two probability measures $\mathbb{P}$ and $\hat{\mathbb{P}}$ on a discrete probability space $\Omega$ is given by $\|\mathbb{P} - \hat{\mathbb{P}}\|_{TV} = \sum_{\omega \in \Omega} (\mathbb{P}(\omega) - \hat{\mathbb{P}}(\omega))_+.$

**Theorem 2.7** ([4, Theorem 2.2]). Let $b = (b(n))_n$ be a sequence of integers with $b(n) = o(\alpha(n)(\log n)^{-1})$. Let $\mathbb{P}_{n,b(n),\alpha}$ be the distribution of $(C_1, \ldots, C_{b(n)})$ under the uniform measure on $S_{n,\alpha}$, and let $\hat{\mathbb{P}}_{b(n)}$ be the distribution of independent Poisson-distributed random variables $(Z_1, \ldots, Z_{b(n)})$ with $\mathbb{E}_{b(n)} (Z_j) = \frac{1}{j}$ for all $j \leq b(n)$. Then there exists $c < \infty$ so that for all $n \in \mathbb{N}$, we have

$$\|\mathbb{P}_{n,b(n),\alpha} - \hat{\mathbb{P}}_{b(n)}\|_{TV} \leq c \left( \frac{\alpha(n)}{n} + b(n) \frac{\log n}{\alpha(n)} \right).$$

In the special case $\alpha(n) \geq \sqrt{n \log(n)}$, Judkovich [16] has computed the above total variation distance using Steins method and obtained a slightly better upper bound.

On the full symmetric group $S_n$, a similar result as Theorem 2.7 holds with $b(n) = o(n)$, see [2]. A natural question at this point is thus if one can replace $b(n)$ in Theorem 2.7 by $b(n) = o(\alpha(n))$. Recall, we have seen in equation (2.8) that

$$\mathbb{E}_{n,\alpha} [C_m(n)] \sim \frac{\alpha^m}{\alpha_m} n \quad \text{as } n \to \infty.$$

Using Lemma 2.1, we immediately see that $\mathbb{E}_{n,\alpha} [C_m(n)] \sim \mathbb{E} Z_m$ if and only if $m = o(\alpha(n)(\log n)^{-1})$. Thus $b(n) = o(\alpha(n)(\log n)^{-1})$ is the most one can expect in Theorem 2.7.

To overcome the problem with the expectations, we replace the random variables $Z_j$ with fixed expectation by random variables $Y_j^{(n)}$ with an expectation depending on $n$ so that

$$\mathbb{E}_{n,\alpha} [C_m(n)] \sim \mathbb{E} Y_m^{(n)} \quad \text{for all } m = o(\alpha(n)).$$

(2.17)
However, to simplify the notation, we write $Y_j$ instead of $Y_j^{(n)}$. We now have

**Theorem 2.8.** Let $b = (b(n))_n$ be a sequence of integers with $b(n) = o(\alpha(n))$. Let $\mathbb{P}_{n, \vartheta, b(n), \alpha}$ be the distribution of $(C_1, \ldots, C_{b(n)})$ under $\mathbb{P}_{n, \alpha}$ on $S_{n, \alpha}$. Further, let $\hat{\mathbb{P}}_{b(n)}$ be the distribution of independent Poisson-distributed random variables $(Y_1, \ldots, Y_{b(n)})$ with $\mathbb{E}[Y_j] = \mu_j(n)$ for all $j \leq b(n)$ and $\mu_j(n)$ as in (2.7). Then

$$\|\mathbb{P}_{n, \vartheta, b(n), \alpha} - \hat{\mathbb{P}}_{b(n)}\|_{TV} = O\left(n^\epsilon \left(\frac{\alpha(n)}{n}\right)^{\frac{5}{12}}\right),$$

where $\epsilon > 0$ is arbitrary. Further, if $b(n) = o(\alpha(n)(\log n)^{-1})$ then

$$\|\mathbb{P}_{n, \vartheta, b(n), \alpha} - \hat{\mathbb{P}}_{b(n)}\|_{TV} = O\left(\frac{\alpha(n)}{n} + \frac{b(n) \log n}{n^{\frac{7}{12}} \alpha + \frac{5}{12}}\right).$$

The proof of this theorem is given in Section 3.5.

### 2.6. Central Limit Theorem for Cycle Numbers

For the proof of Proposition 2.2, we require a central limit theorem for the cycle counts in the case $\mathbb{E}[C_m] \to \infty$. The main result of this section is to establish this theorem. Explicitly, we prove the following.

**Theorem 2.9.** Let $m_k : \mathbb{N} \to \mathbb{N}$ for $1 \leq k \leq K$ such that $m_k(n) \leq \alpha(n)$ and $m_{k_1}(n) \neq m_{k_2}(n)$ if $k_1 \neq k_2$ for large $n$. Suppose that

$$\mu_{m_k(n)}(n) \to \infty$$

for all $k$. We then have as $n \to \infty$

$$\left(\frac{C_{m_1(n)} - \mu_{m_1(n)}(n)}{\sqrt{\mu_{m_1(n)}(n)}} , \ldots , \frac{C_{m_K(n)} - \mu_{m_K(n)}(n)}{\sqrt{\mu_{m_K(n)}(n)}}\right) \xrightarrow{d} (N_1, \ldots, N_K),$$

with $N_1, \ldots, N_K$ independent standard normal distributed random variables.

This theorem was proven in [4] under the additional assumption

$$n^{-\frac{5}{12}} \alpha(n)^{-\frac{7}{12}} \frac{x_{n, \vartheta}}{\sqrt{\mu_{m_k(n)}(n)}} \to 0.$$  

(2.20)

In Section 3.6 we present a proof that does not require the additional assumption.
3. Proofs

3.1. Generating functions and the saddle point method

Generating functions and their connection with analytic combinatorics form the backbone of the proofs in this paper. More precisely, we will determine formal generating functions for all relevant moment-generating functions and then use the saddle-point method to determine the asymptotic behaviour of these moment-generating functions as \( n \to \infty \).

Let \((a_n)_{n \in \mathbb{N}}\) be a sequence of complex numbers. Then its ordinary generating function is defined as the formal power series

\[
    f(z) := \sum_{n=0}^{\infty} a_n z^n.
\]

The sequence may be recovered by formally extracting the coefficients

\[
    [z^n] f(z) := a_n
\]

for any \( n \). The first step is now to consider a special case of Polya’s Enumeration Theorem, see [21, §16, p. 17], which connects permutations with a specific generating function.

**Lemma 3.1.** Let \((q_j)_{j \in \mathbb{N}}\) be a sequence of complex numbers. We then have the following identity between formal power series in \( z \),

\[
    \exp \left( \sum_{j=1}^{\infty} \frac{q_j z^j}{j} \right) = \sum_{k=0}^{\infty} \frac{z^k}{k!} \sum_{\sigma \in S_k} \prod_{j=1}^{k} q_j^{C_j}, \tag{3.1}
\]

where \( C_j = C_j(\sigma) \) are the cycle counts. If either of the series in (3.1) is absolutely convergent, then so is the other one.

Extracting the \( n \)th coefficient yields

\[
    [z^n] \exp \left( \sum_{j=1}^{\infty} \frac{q_j z^j}{j} \right) = \frac{1}{n!} \sum_{\sigma \in S_n} \prod_{j=1}^{n} q_j^{C_j}. \tag{3.2}
\]

With this formulation, the parameters \((q_j)\) can depend on the system size \( n \). For instance, setting \( q_j = \vartheta 1_{(j \leq \alpha(n))} \), we obtain

\[
    Z_{n,\alpha} = [z^n] \exp \left( \vartheta \sum_{j=1}^{\alpha(n)} \frac{z^j}{j} \right). \tag{3.3}
\]
with $Z_{n,\alpha}$ as in (2.4). Similarly, we can get an expression for the moment generating function of $C_{m(n)}$, where $(m(n))_{n\in \mathbb{N}}$ is an integer sequence with $m(n) \leq \alpha(n)$. Indeed, setting $q_{m(n)} = \vartheta e^x$ and $q_j = \vartheta \mathbb{1}_{\{j \leq \alpha(n)\}}$ for $j \neq m(n)$, we get

$$E_n,\alpha[ e^{x C_{m(n)}}] = \frac{1}{Z_{n,\alpha}} [z^n] \exp \left( \vartheta (e^x - 1) \frac{z^{m(n)}}{m(n)} \right) \exp \left( \vartheta \sum_{j=1}^{\alpha(n)} \frac{z^j}{j} \right).$$  

(3.4)

In view of (3.3) and (3.53), we can compute the asymptotic behaviour of $Z_{n,\alpha}$ (and similar expressions) by extracting the coefficients of power series as in (3.3) and (3.53). One way to extract these coefficients is the saddle point method, a standard tool in asymptotic analysis. The basic idea is to rewrite the expression (3.2) as a complex contour integral and choose the path of integration in a convenient way. The details of this procedure depend on the situation at hand and need to be done on a case by case basis. A general overview over the saddle-point method can be found in [15, page 551]. An important part of this computations is typically to find a solution of the so-called saddle-point equation.

We now treat the most general case of the saddle point method that is relevant for the present situation. Let $q = (q_{j,n})_{1 \leq j \leq \alpha(n), n \in \mathbb{N}}$ be a triangular array. We assume that all $q_{j,n}$ are nonnegative, real numbers and that for each $n \in \mathbb{N}$ there exists a $j$ such that $q_{j,n} > 0$. We then define $x_{n,q}$ as the unique positive solution of

$$n = \sum_{j=1}^{\alpha(n)} q_{j,n} x_{n,q}^j.$$  

(3.5)

Let further

$$\lambda_{p,n} := \lambda_{p,n,\alpha,q} := \sum_{j=1}^{\alpha(n)} q_{j,n} j^{p-1} x_{n,q}^j,$$  

(3.6)

where $p \geq 0$ is a natural number. Due to Equation (3.5),

$$\lambda_{p,n} \leq n (\alpha(n))^{p-1}.$$  

(3.7)

holds for all $p \geq 1$. Note that (3.7) may not hold for $p = 0$. We now define

\textbf{Definition 3.2.} A triangular array $q$ is called admissible if the following three conditions are satisfied:

1. It satisfies

$$\alpha(n) \log(x_{n,q}) \approx \log \left( \frac{n}{\alpha(n)} \right).$$

2. We have

$$\lambda_{2,n} \approx n \alpha(n).$$
3. There exist a non-negative sequence \((b(n))_{n \in \mathbb{N}}\) and constants \(\delta, c > 0\) such that \(b(n)/\alpha(n) < 1 - \delta\) and \(q_{j,n} \geq c > 0\) for all \(j \geq b(n)\) hold for \(n\) large enough.

Note that condition (1) implies in particular that \(x_{n,q} > 1\) and that \(x_{n,q} \to 1\) as \(n \to \infty\). Let \(B_r(0)\) denote the ball with center 0 and radius \(r\) in the complex plane.

**Definition 3.3.** Let \(q\) be an admissible triangular array. Then a sequence \((f_n)_{n \in \mathbb{N}}\) of functions is called admissible (w.r.t. \(q\)) if it satisfies the following three conditions:

1. There is \(\delta > 0\) such that \(f_n\) is holomorphic on the disc \(B_{x_{n,q} + \delta}(0)\) if \(n \in \mathbb{N}\) is large enough.
2. There exist constants \(K, N > 0\) such that
   \[
   \sup_{z \in \partial B_{x_{n,q}}(0)} |f_n(z)| \leq n^K |f_n(x_{n,q})|
   \]
   for all \(n \geq N\).
3. With the definition
   \[
   \|f_n\|_n := n^{-\frac{\alpha(n)}{2}} \sup_{|\varphi| \leq n^{-\frac{\alpha(n)}{2}}} \frac{|f_n(x_{n,q} e^{i\varphi})|}{|f_n(x_{n,q})|},
   \]
   we have \(\|f_n\|_n \to 0\) as \(n \to \infty\).

We are now in the position to formulate our general saddle point result.

**Proposition 3.4 ([4, Proposition 3.2]).** Let \(q\) be an admissible triangular array and \((f_n)_{n \in \mathbb{N}}\) an admissible sequence of functions. Then we have as \(n \to \infty\)

\[
[z^n] f_n(z) \exp \left( \sum_{j=1}^{\alpha(n)} q_{j,n} \frac{z^j}{j} \right) = \frac{f_n(x_{n,q}) e^{\lambda_{0,n}}}{x_{n,q}^{1/\sqrt{2\pi \lambda_{2,n}}}} \left( 1 + O \left( \frac{\alpha(n)}{n} + \|f_n\|_n \right) \right).
\]

Note that the implicit constants in the \(O(,\) terms in (3.10) can depend on \(K, N\) and \(\delta\) from the above definition of admissibility. However, we require for our computations only the leading term in (3.10). Also we will not vary the values of \(K, N\) and \(\delta\). Thus we need only the existence of \(K, N\) and \(\delta\), but not their values. We therefore can safely omit the dependence on \(K, N\) and \(\delta\).

In view of Proposition 3.4, we see it is important to understand the asymptotic behaviour of \(x_{n,q}\) and \(\lambda_{j,n}\) as \(n \to \infty\). Lemma 2.1 will be very useful for this purpose.

### 3.2. Proof of Proposition 2.2

We have by assumption \(\mu_{\alpha(n)} \to \infty\). Thus we can apply Theorem 2.9. We conclude that

\[
\frac{C_{\alpha(n)} - \mu_{\alpha(n)}}{\sqrt{\mu_{\alpha(n)}(n)}} \xrightarrow{d} N,
\]

\[(3.11)\]
where $N$ is a standard normal distributed random variable. Since $\alpha(n)$ is the maximal cycle length, we have

$$
\mathbb{P}_{n,\alpha}[((\ell_1, \ell_2, \ldots, \ell_K) \neq (\alpha(n), \alpha(n), \ldots, \alpha(n)))] = \mathbb{P}_{n,\alpha}[C_{\alpha(n)} < K].
$$

Using (3.11), we get

$$
\mathbb{P}_{n,\alpha}[C_{\alpha(n)} < K] = \mathbb{P}_{n,\alpha}\left[\frac{C_{\alpha(n)} - \mu_{\alpha(n)}}{\sqrt{\mu_{\alpha(n)}}} < \frac{K - \mu_{\alpha(n)}}{\sqrt{\mu_{\alpha(n)}}}\right] \xrightarrow{n \to \infty} 0,
$$

and the claim follows.

### 3.3. Proof of Proposition 2.3

As a first step, we state

**Proposition 3.5.** Let $(m_k(n))_{n \in \mathbb{N}}$, $k = 1, \ldots, d$, be integer sequences satisfying $1 \leq m_k(n) \leq \alpha(n)$ and $m_k(n) \neq m_\ell(n)$ for $k \neq \ell$. Suppose that

$$
\mu_{m_k(n)} \to \mu_k \in [0, \infty[ \quad \text{for all } k.
$$

Then

$$
(C_{m_1(n)}, \ldots, C_{m_d(n)}) \xrightarrow{d} (Y_1, \ldots, Y_d) \quad (3.12)
$$

where $(Y_k)_{k=1}^d$ a sequence of independent Poisson distributed random variables with

$$
\mathbb{E}[Y_k] = \mu_k \text{ for all } k = 1, \ldots, d.
$$

This proposition was proven in [4], but we give the proof of this proposition for the case $d = 1$ for the convenience of the reader.

**Proof.** Let $d = 1$. We argue here with the moment generating function. We saw in (3.53) that we have for $s \geq 0$

$$
\mathbb{E}_{n,\alpha}[e^{sC_{m_1(n)}}] = \frac{1}{Z_{n,\alpha}} [z^n] \exp \left( \vartheta (e^s - 1) z_{m_1(n)}^{m_1(n)} \right) \exp \left( \vartheta \sum_{j=1}^{\alpha(n)} \frac{z_j}{j} \right). \quad (3.13)
$$

We now apply Proposition 3.4 to compute the asymptotic behaviour of this expression in the case $s \geq 0$. According to [28], this is sufficient to prove the proposition. We use $q = (q_{j,n})$ with $q_j = \vartheta \mathbb{1}_{(j \leq \alpha(n))}$ and $f_n(z) = \exp \left( \vartheta (e^s - 1) z_{m_1(n)}^{m_1(n)} \right)$. We thus have to show that $q$ and the sequence $(f_n(n))_{n \in \mathbb{N}}$ are admissible, see Definitions 3.2 and 3.3. Inserting the definition of $q$, we immediately get that the corresponding saddle point equation is given by (2.6), hence the solution is $x_{n,\alpha}$. The admissibility of $q$ then follows.
immediately from Lemma 2.1. It remains to show that \((f_n)_{n \in \mathbb{N}}\) is admissible. All \(f_n\) are entire functions and hence we can choose any \(\delta > 0\). Since \(s \geq 0\), we have for all \(r > 0\) and \(\varphi \in [-\pi, \pi] \]

\[
|f_n(re^{i\varphi})| \leq |f_n(r)|.
\]

Thus the second condition is fulfilled with \(K = 0\). For the third condition, we use

\[
f'_n(z) = \vartheta(e^s - 1)z^{m_1(n)-1}f_n(z)
\]

and that \(\mu_{m_1(n)} = \mathbb{E}_{m_1(n)}^{\mu_{m_1(n)}}\). Inserting this and that \(\mu_{m_1(n)} \to \mu_1\) immediately shows that the third condition is fulfilled. So we can apply Proposition 3.4. Using that \(\mathbb{E}_{n,\alpha}[e^{e^{C_{m_1(n)}}}] = 1\) for \(s = 0\), we obtain

\[
\mathbb{E}_{n,\alpha}[e^{e^{C_{m_1(n)}}}] \to \exp(\vartheta(e^s - 1)\mu_1).
\]

(3.14)

This completes the proof. \(\square\)

Now we turn to the proof of Proposition 2.3. In this proof, we write \(\mu_{\alpha(n)}(n)\) instead of \(\mu_{\alpha(n)}\). Let \(j \in \mathbb{N}_0\) be arbitrary. Using the definition of \(\mu_m(n)\) in (2.7) with \(m = \alpha(n) - j\), we get

\[
\frac{\mu_{\alpha(n)}(n)}{\mu_{\alpha(n) - j}(n)} = \frac{\alpha(n) - j}{\alpha(n)} \chi_{n,\alpha}^{\alpha(n) - j} \xrightarrow{n \to \infty} 0.
\]

Since \(\mu_{\alpha(n)}(n) \to \mu\) by assumption, we get that

\[
\mu_{\alpha(n) - j}(n) \xrightarrow{n \to \infty} \mu
\]

for all \(j \in \mathbb{N}_0\). Proposition 3.5 therefore implies that the cycle counts \((C_{\alpha(n) - j})_{0 \leq j \leq d}\) converge in distribution to a sequence \((Z_j)_{j=0}^d\), where \((Z_j)_{j=0}^d\) is i.i.d. Poisson distributed with parameter \(\mu\). We now have, as \(n \to \infty\),

\[
\mathbb{P}_{n,\alpha}[\ell_k \leq \alpha(n) - d] = \mathbb{P}_{n,\alpha}\left[\sum_{i=0}^{d-1} C_{\alpha(n) - i} \leq k - 1\right] \to \mathbb{P}\left[\sum_{i=0}^{d-1} Z_i \leq k - 1\right].
\]

By the independence of \((Z_j)_{0 \leq j \leq d}\), the random variable \(\sum_{i=0}^{d-1} Z_i\) is Poisson-distributed with parameter \(d\mu\). Thus,

\[
\mathbb{P}\left[\sum_{j=0}^{d-1} Z_j \leq k - 1\right] = \sum_{j=0}^{k-1} \frac{e^{-d\mu}(d\mu)^j}{j!} = \frac{1}{\Gamma(k)} \int_{d\mu}^{\infty} v^{k-1}e^{-v}\text{d}v,
\]

where \(\Gamma(s)\) denotes the gamma function. The last equality follows by partial integration and induction. We now have

\[
\mathbb{P}_{n,\alpha}[\ell_k = \alpha(n) - d] = \mathbb{P}_{n,\alpha}[\ell_k \leq \alpha(n) - d] - \mathbb{P}_{n,\alpha}[\ell_k \leq \alpha(n) - (d + 1)].
\]
This implies
\[ P_{n,\alpha} [\ell_k = \alpha(n) - d] \xrightarrow{n \to \infty} \frac{1}{\Gamma(k)} \int_{d}^{(d+1)\mu} v^{k-1}e^{-v}dv. \]
The claim is proved.

**Remark 3.6.** The proof of Proposition 2.3 can also be used to compute the limit of
\[ P_{n,\alpha} \left( (\ell_k)_{k=1}^{d} = (\alpha(n) - d_{k})_{k=1}^{d} \right) \]
as \( n \) tends to infinity since the event in question only depends on a finite number of cycle counts \( C_{\alpha(n)-j} \). It is, however, cumbersome to provide a closed form for such probabilities:

The reason for this is that the stochastic process \((\ell_k)_{k=1}^{d}\) is not Markovian, i.e. the distribution of \( \ell_d \) depends non-trivially on the distribution of the random vector \((\ell_k)_{k=1}^{d-1}\).

This is why we only provide the readily interpretable results for one individual \( \ell_k \) at a time in the proposition.

### 3.4. Proof of Theorem 2.5

We will first prove certain auxiliary results, assuming that \( \mu_{\alpha(n)} \to 0 \). Inserting the definition of \( \mu_{\alpha(n)} \), see (2.7), we get
\[ \mu_{d_t(n)}(n) = \frac{d_t(n)}{d_t(n) + \mu_{\alpha(n)}} = \frac{x_{n,\alpha}(n) - \lfloor t/\mu_{\alpha(n)} \rfloor}{\alpha(n) - \lfloor t/\mu_{\alpha(n)} \rfloor} = \frac{\alpha(n)}{\alpha(n) - \lfloor t/\mu_{\alpha(n)} \rfloor} x_{n,\alpha}(n) - \lfloor t/\mu_{\alpha(n)} \rfloor. \]

We now have
\[ \frac{\alpha(n)}{\alpha(n) - \lfloor t/\mu_{\alpha(n)} \rfloor} \xrightarrow{n \to \infty} 1, \]
locally uniformly in \( t \) since \( 1/\mu_{\alpha(n)} = o(\alpha(n)) \) by Equation (2.14). By Lemma 2.1 and Equation (2.14), we have as \( n \to \infty \)
\[ x_{n,\alpha}^{-\lfloor t/\mu_{\alpha(n)} \rfloor} = \exp \left( -\frac{t}{\mu_{\alpha(n)}} \frac{1}{\alpha(n)} \log \left( \frac{n}{\partial \alpha(n)} \log \left( \frac{n}{\partial \alpha(n)} \right) \right) (1 + o(1)) \right) \]
\[ = \exp \left( \mathcal{O} \left( \frac{\alpha(n)}{n} \right) \right) \xrightarrow{t \geq 0} 1, \quad (3.15) \]
locally uniformly in \( t \geq 0 \). Altogether, we have locally uniformly in \( t \) that
\[ \mu_{d_t(n)}(n) \sim \mu_{\alpha(n)}. \]
Furthermore, the function \( m \to \mu_{m}(n) \) is increasing for \( m \geq \frac{\alpha(n)}{\mu_{\alpha(n)}} \). This follows by computing the derivative with respect to \( m \) of \( \mu_{m}(n) \) in (2.7) and using Lemma 2.1. We thus have locally uniformly in \( t \)
\[ \sum_{m=d_t(n)+1}^{\alpha(n)} \mu_{m}(n) \xrightarrow{n \to \infty} \frac{t}{\mu_{\alpha(n)}} \mu_{\alpha(n)} = t. \quad (3.16) \]
In order to establish convergence as a stochastic process, we begin by proving convergence of the finite-dimensional distributions. More precisely, for $0 = t_0 < t_1 < \ldots < t_K$ and $K \in \mathbb{N}$, consider the increments $(P_{t_k} - P_{t_{k-1}})_{k=1}^K$. We now have

$$P_{t_k} - P_{t_{k-1}} = \sum_{j=d_{t_k}(n)+1}^{d_{t_{k-1}}(n)} C_j.$$  \hfill (3.17)

We begin by determining the moment generating function. We have

$$\mathbb{E}_{n,\alpha} \left[ \prod_{k=1}^{K} \exp \left( s_k (P_{t_k} - P_{t_{k-1}}) \right) \right] = \mathbb{E}_{n,\alpha} \left[ \exp \left( \sum_{k=1}^{K} e^{s_k} \sum_{j=d_{t_k}+1}^{d_{t_{k-1}}} C_j \right) \right]$$

$$= \frac{1}{Z_{n,\alpha,\delta}} [z^n] \exp \left( \sum_{k=1}^{K} (e^{s_k} - 1) \sum_{j=d_{t_k}+1}^{d_{t_{k-1}}} \frac{\vartheta}{j} z^j \right) \exp \left( \sum_{j=1}^{\alpha(n)} \frac{\vartheta}{j} z^j \right),$$  \hfill (3.18)

where $s_k \geq 0$ for all $1 \leq k \leq K$. Equation (3.18) follows immediately with Lemma 3.1 and a small computation using $q_j = \vartheta e^{s_k}$ for $d_{t_k} < j \leq d_{t_{k-1}}$.

We will apply Proposition 3.4 with $q = (q_j)_{n}$ with $q_j = \vartheta {1_{j \leq \alpha(n)}}$ and the perturbations

$$f_{n}(z) = \exp \left( \sum_{k=1}^{K} (e^{s_k} - 1) \sum_{j=d_{t_k}+1}^{d_{t_{k-1}}} \frac{\vartheta}{j} z^j \right).$$

To do this, we have to check that the array $q$ and the sequence $(f_{n})_{n \in \mathbb{N}}$ are admissible, see Definitions 3.2 and 3.3. The array $q$ is admissible by Lemma 2.1. Let us now look at $(f_{n})_{n \in \mathbb{N}}$. The functions $f_{n}$ are entire. Thus we can use any $\delta > 0$. Further, all coefficients of the Taylor expansion of $f_{n}(z)$ at $z = 0$ are non-negative since all $s_k \geq 0$. This implies

$$|f_{n}(z)| \leq f_{n}(x_{n,\delta}) \text{ for all } z \in \mathbb{C} \text{ with } |z| = x_{n,\delta}.$$  

It remains to check condition (3.9). We have

$$f'_{n}(z) = \sum_{k=1}^{K} (e^{s_k} - 1) \sum_{j=d_{t_k}+1}^{d_{t_{k-1}}} \vartheta z^{j-1} f_{n}(z).$$

We thus have for all $z \in \mathbb{C}$ with $|z| = x_{n,\delta}$ that

$$\left| \frac{f'_{n}(z)}{f_{n}(x_{n,\delta})} \right| \leq \sum_{k=1}^{K} (e^{s_k} - 1) \sum_{j=d_{t_k}+1}^{d_{t_{k-1}}} \vartheta x_{n,\delta}^{j-1}$$

$$\leq \vartheta \sum_{j=1}^{\alpha(n)} \frac{(e^{s_k} - 1) \sum_{j=d_{t_k}+1}^{d_{t_{k-1}}} \vartheta x_{n,\delta}^{j-1}}{\sum_{j=1}^{\alpha(n)} \frac{(e^{s_k} - 1) \sum_{j=d_{t_k}+1}^{d_{t_{k-1}}} \vartheta x_{n,\delta}^{j-1}}{1}.}$$  \hfill (3.19)
Using the definition of $\mu_{\alpha(n)}$ in (2.7), we see that we have locally uniformly in $s_k$
\[
\left| \frac{f'_{n,s} (z)}{f_{n,s} (x_{n}, \theta)} \right| = \mathcal{O} \left( \varphi_{x_{n} \theta} \frac{t_{k}}{\mu_{\alpha(n)}} \right) = \mathcal{O} \left( \alpha(n) \right). 
\]
Inserting this into (3.9), we obtain
\[
\|f_{n,s}\|_{n} = n^{-\frac{1}{12}} (\alpha(n))^{-\frac{7}{12}} \sup_{|\varphi| \leq n^{-\frac{1}{12}} (\alpha(n))^{-\frac{1}{2}}} \left| \frac{f'_{n,s}(x_{n,q}^{\varphi})}{f_{n,s}(x_{n,q})} \right| 
\leq n^{-\frac{1}{12}} (\alpha(n))^{-\frac{7}{12}} \mathcal{O}(\alpha(n)) = \mathcal{O} \left( \frac{\alpha(n)}{n} \right)^{5/12} \to 0.
\]
This implies that the sequence $(f_{n,s})_{n \in \mathbb{N}}$ is admissible, so we can apply Proposition 3.4 to (3.18). Observe that Equation (3.16) entails
\[
\sum_{j=d_{k-1}}^{\mathcal{d}_{k-1}} \mu_{j}(n) = \alpha(n) - \frac{t_{k-1}}{\mu_{n}} \sum_{j=\alpha(n) - \frac{t_{k}}{\mu_{n}} + 1}^{\mathcal{d}_{k-1}} \mu_{j}(n) \xrightarrow{n \to \infty} t_{k} - t_{k-1}
\]
for all $k$. Since we use for all $s_k$ the same array $q$, including the case $s_{1} = \ldots = s_{K} = 0$, we get with Proposition 3.4 that
\[
\mathbb{E}_{n,\alpha} \left[ \prod_{k=1}^{K} \exp \left( s_{k} \left( P_{t_{k}} - P_{t_{k-1}} \right) \right) \right] \sim f_{n,s}(x_{n}, \theta) = \exp \left( \sum_{k=1}^{K} (e^{s_{k}} - 1) \sum_{j=d_{k-1}+1}^{\mathcal{d}_{k-1}} \mu_{j}(n) \right) \xrightarrow{k \to 1} \sum_{k=1}^{K} \exp \left( (e^{s_{k}} - 1)(t_{k} - t_{k-1}) \right).
\]
This implies that the increments $(P_{t_{k}} - P_{t_{k-1}})_{k=1}^{K}$ converge in distribution to independent random variables $(Z_{1}, Z_{2}, \ldots, Z_{K})$, where $Z_{k}$ is Poisson-distributed with parameter $t_{k} - t_{k-1}$. Thus the finite-dimensional distributions of $P_{t}$ converge weakly to the finite-dimensional distributions of the Poisson process with parameter $1$.

To prove that the process $\{P_{t}, t \geq 0\}$ converges to the Poisson process with parameter 1, it remains to establish the tightness of the process $\{P_{t}, t \geq 0\}$. By [9, Theorem 13.5 and (13.14)], it is sufficient to show for each $T > 0$ that
\[
\mathbb{E}_{n,\alpha} \left[ (P_{t} - P_{t_{1}})^{2} (P_{t_{2}} - P_{t})^{2} \right] = \mathcal{O} \left( (t_{2} - t_{1})^{2} \right) \quad (3.20)
\]
uniformly in $t, t_{1}, t_{2}$ with $0 \leq t_{1} \leq t \leq t_{2} \leq T$. Note that we can assume that $\frac{t_{2}}{\mu_{n}(n)} - \frac{t_{1}}{\mu_{n}(n)} \geq 1$. Otherwise $(P_{t_{1}} - P_{t_{1}})^{2} (P_{t_{2}} - P_{t})^{2} = 0$ and the above equation is trivially fulfilled. Let $n$ be large enough such that $d_{T}(n) > 0$. By Equation (3.18), we
have
\[
E_{n,\alpha} \left[ (P_t - P_{1,t})^2 (P_{t_2} - P_{1})^2 \right] = \frac{\partial^2}{\partial s_t^2 \partial s_1^2} E_{n,\alpha} \left[ e^{s_1 (P_{1,t} - P_{1,t}) + s_2 (P_{2} - P_{1})} \right]_{s_t = s_1 = 0}
\]
\[
= \frac{1}{Z_{n,\alpha,\vartheta}} \frac{\partial^2}{\partial s_t^2 \partial s_1^2} \left[ z^n \exp \left( (e^{s_1} - 1) G_{n,t_1,t} (z) + (e^{s_2} - 1) G_{n,t,t_2} (z) \right) \right]_{s_t = s_1 = 0}
\]
with \( G_{n,u,w} (z) := \sum_{j=d_u(n)} \vartheta_j z^j \) for \( 0 \leq u \leq w \leq T \). Calculating the derivatives and entering \( s_1 = s_2 = 0 \) gives
\[
E_{n,\alpha} \left[ (P_t - P_{1,t})^2 (P_{t_2} - P_{1})^2 \right] = \frac{1}{Z_{n,\alpha,\vartheta}} [z^n] g_n (z) \exp \left( \sum_{j=1}^{\alpha(n)} \frac{\vartheta_j}{j} z^j \right)
\]
with
\[
g_n (z) := G_{n,t_1,t} (z) (1 + G_{n,t_1,t} (z)) G_{n,t,t_2} (z) (1 + G_{n,t_2,t} (z)) \).
\]

We now apply again Proposition 3.4. We use here the perturbations \((g_n)_{n \in \mathbb{N}}\) and as before \( q = (q_j,n) \) with \( q_j = \vartheta \mathbb{1}_{(j \leq \alpha(n))} \). Thus we only have to show that \((g_n)_{n \in \mathbb{N}}\) is admissible. All \( g_n \) are entire and we thus can use any \( \delta > 0 \). Further the coefficients of the Taylor expansion of \( g_n(z) \) at \( z = 0 \) are all non-negative. Thus \( |g_n(z)| \leq g_n(|z|) \) for all \( z \). It remains to check condition (3.9). We use here an estimate which is similar to the one in (3.19). We have for \( z \in \mathbb{C} \) with \( |z| = x_{n,\vartheta} \) that
\[
|G'_{n,u,w} (z)| = \left| \vartheta \sum_{j=d_u(n)+1}^{d_u(n)} \vartheta_j z^{j-1} \right| \leq \vartheta \sum_{j=d_u(n)+1}^{d_u(n)} x_{n,\vartheta}^{j-1} \leq \vartheta x_{n,\vartheta}^{\alpha(n)-1} \sum_{j=\alpha(n)-1}^{\alpha(n)} \left| \frac{w}{\mu_\alpha(n)} \right| + 1
\]
\[
= \vartheta x_{n,\vartheta}^{\alpha(n)} \left( \left| \frac{w}{\mu_\alpha(n)} \right| - \left| \frac{u}{\mu_\alpha(n)} \right| \right).
\]
Similarly, we have
\[
|G_{n,u,w} (x_{n,\vartheta})| = \vartheta \sum_{j=d_u(n)+1}^{d_u(n)} x_{n,\vartheta}^{j-1} \geq \frac{\vartheta}{d_u(n)} x_{n,\vartheta}^{d_u(n)+1} \sum_{j=d_u(n)+1}^{d_u(n)} 1 \geq \frac{\vartheta}{d_u(n)} x_{n,\vartheta}^{d_u(n)+1} \left( \left| \frac{w}{\mu_\alpha(n)} \right| - \left| \frac{u}{\mu_\alpha(n)} \right| \right). \tag{3.21}
\]
Using (3.15) and the definition of \( d_u(n) \) in (2.16), we get
\[
\left| \frac{G'_{n,u,w} (z)}{G_{n,u,w} (x_{n,\vartheta})} \right| \leq d_u(n) x_{n,\vartheta}^{\alpha(n)+1} \leq \alpha(n) \exp \left( \mathcal{O} \left( \frac{T^{\alpha(n)}}{n} \right) \right) = \mathcal{O} (\alpha(n)).
\]
This estimate is uniform in \( u, w \) with \( 0 \leq u \leq w \leq T \). Inserting this inequality into (3.9) then gives

\[
\|G_{n,u,w}\|_n \leq n^{-\frac{5}{12}} (\alpha(n))^{-\frac{7}{12}} O(\alpha(n)) = O\left(\left(\frac{\alpha(n)}{n}\right)^{5/12}\right) \rightarrow 0.
\]

We thus have

\[
\|g_n\|_n \leq 2\|G_{n,t_1,t}\|_n + 2\|G_{n,t_2,t}\|_n = O\left(\left(\frac{\alpha(n)}{n}\right)^{5/12}\right). \tag{3.22}
\]

This estimate is uniform in \( t, t_1, t_2 \) with \( 0 \leq t_1 \leq t \leq t_2 \leq T \). This implies that the sequence \((g_n)_{n \in \mathbb{N}}\) is admissible. Proposition 3.4 then implies that

\[
E_{n,\alpha}\left[(P_t - P_{t_1})^2 (P_{t_2} - P_t)^2\right] = g_n(x_n,\vartheta) \left(1 + O\left(\frac{\alpha(n)}{n} + \|g_n\|_n\right)\right) \leq 2g_n(x_n,\vartheta).
\]

Using the definition of \( g_n \) and an estimate similar to (3.21), we get

\[
g_n(x_n,\vartheta) \leq \left(\sum_{j=d_{t_2}(n)+1}^{d_{t_1}(n)} \frac{\vartheta_j^2}{x_{n,\vartheta}^j}\right)^2 \left(1 + \sum_{j=d_{t_2}(n)+1}^{d_{t_1}(n)} \frac{\vartheta_j^2}{x_{n,\vartheta}^j}\right)^2
\]

\[
\leq 2(d_{t_1}(n) - d_{t_2}(n))^2 \mu_{\alpha(n)}^2 \left(1 + 2(d_{t_1}(n) - d_{t_2}(n)) \mu_{\alpha(n)}\right)^2.
\]

Using the definition of \( d_t(n) \) in (2.16) and that \( 0 \leq t_1 \leq t_2 \leq T \), we obtain

\[
g_n(x_n,\vartheta) \leq 2(1 + 2T)^2 (d_{t_1}(n) - d_{t_2}(n))^2 \mu_{\alpha(n)}^2
\]

\[
= 2(1 + 2T)^2 \left(\frac{t_2}{\mu_{\alpha(n)}} - \left\lfloor \frac{t_1}{\mu_{\alpha(n)}} \right\rfloor\right)^2 \mu_{\alpha(n)}^2
\]

\[
\leq 2(1 + 2T)^2 \left(\frac{t_2}{\mu_{\alpha(n)}} - \frac{t_1}{\mu_{\alpha(n)}} + 1\right)^2 \mu_{\alpha(n)}^2
\]

\[
\leq 8(1 + 2T)^2 (t_2 - t_1)^2.
\]

Note that we used for the last equation the assumption \( \frac{t_2}{\mu_{\alpha(n)}} - \frac{t_1}{\mu_{\alpha(n)}} \geq 1 \). This shows that (3.20) holds. This completes the proof.

### 3.5. Proof of Theorem 2.8

The proof follows mainly the ideas in [2], where the case of uniform permutations is treated, and is also similar to the proof of Theorem 2.7 in [4].

In order to establish Theorem 2.8, we have to introduce some notation. We set

\[
d_{b(n)} := \|P_{n,\vartheta,\beta(n),\alpha} - \hat{P}_{b(n)}\|_{\text{TV}}. \tag{3.23}
\]
Let \( (Y_j) \) be as in Theorem 2.8 and set for \( b_1, b_2 \in \mathbb{N} \)
\[
T^{(n)}_{b_1 b_2} := \sum_{j=b_1+1}^{b_2} j Y_j .
\]  

(3.24)

Further, let \( C_b = (C_1, C_2, \ldots, C_{b(n)}) \) the vector of the cycle counts up to length \( b(n) \), \( Y_b = (Y_1, Y_2, \ldots, Y_{b(n)}) \), and \( c = (c_1, c_2, \ldots, c_{b(n)}) \in \mathbb{N}^{b(n)} \) a vector. We then have for all \( c \)
\[
\mathbb{P}_{n,\alpha} [ C_b = c ] = \mathbb{P} \left[ Y_b = c | T^{(n)}_{0\alpha(n)} = n \right] .
\]  

(3.25)

The proof of this equality is the same as for the uniform measure on \( S_n \) in \([2]\) and we thus omit it. As in \([4, \text{Section 4.2}]\), one can use (3.25) to show that
\[
d_{b(n)} = \sum_{r=0}^{\infty} \mathbb{P} \left[ T^{(n)}_{0\alpha(n)} = r \right] \left( 1 - \frac{\mathbb{P} \left[ T^{(n)}_{b(n)\alpha(n)} = n-r \right]}{\mathbb{P} \left[ T^{(n)}_{0\alpha(n)} = n \right]} \right) ,
\]  

(3.26)

where \((y)_+ = \max(y, 0)\). We will split this sum into pieces. We have
\[
d_{b(n)} \leq \mathbb{P} \left[ T^{(n)}_{0\alpha(n)} \geq \rho \mathbb{E} [ T^{(n)}_{0\alpha(n)} ] \right] + \max_{1 \leq r \leq \rho \mathbb{E} [ T^{(n)}_{0\alpha(n)} ]} \left( 1 - \frac{\mathbb{P} \left[ T^{(n)}_{b(n)\alpha(n)} = n-r \right]}{\mathbb{P} \left[ T^{(n)}_{0\alpha(n)} = n \right]} \right) ,
\]

where \( \rho = \rho(n) > 1 \) is arbitrary. We now have

**Lemma 3.7.** Let \( \rho > 1 \). Then,
\[
\mathbb{P} \left[ T^{(n)}_{0\alpha(n)} \geq \rho \mathbb{E} [ T^{(n)}_{0\alpha(n)} ] \right] \leq \exp \left( \mathbb{E} [ T^{(n)}_{0\alpha(n)} ] \rho - \rho \log(\rho) \right) .
\]

**Proof.** We set \( m := \mathbb{E} [ T^{(n)}_{0\alpha(n)} ] \). We then have for all \( s \geq 0 \)
\[
\mathbb{P} \left[ T^{(n)}_{0\alpha(n)} \geq \rho m \right] = \mathbb{P} \left[ e^{s T^{(n)}_{0\alpha(n)}} \geq e^{\rho m} \right] \leq \frac{\mathbb{E} \left[ e^{s T^{(n)}_{0\alpha(n)}} \right]}{e^{\rho m}} .
\]  

(3.27)

The independence of the \( Y_j \) and \( m = \sum_{j=1}^{b(n)} j \mu_j(n) = \theta \sum_{j=1}^{b(n)} x_{n,\theta}^j \) imply that
\[
\log \left( \mathbb{E} \left[ e^{s T^{(n)}_{0\alpha(n)}} \right] \right) = \sum_{j=1}^{b(n)} \mu_j(n) (e^{s} - 1) = \theta \sum_{j=1}^{b(n)} x_{n,\theta}^j \int_0^s e^{sj} dx \leq \theta \sum_{j=1}^{b(n)} x_{n,\theta}^j \int_0^s e^{b(n)x} dx
\]
\[
\leq m \int_0^s e^{b(n)x} dx = m \frac{e^{b(n)s} - 1}{b(n)} \leq m \frac{e^{b(n)s}}{b(n)} .
\]  

(3.28)
We thus have \( P \left[ T_{(n)}^{(n)} \geq \rho m \right] \leq \exp \left( \frac{me^{b(n)}}{b(n)^2} - s \rho m \right) \). We now use \( s = \frac{1}{\log(\rho)} \), which is by assumption non-negative. Inserting this into the above inequality completes the proof.

In order to choose a suitable \( \rho \), we have to determine the asymptotic behavior of \( E \left[ T_{(n)}^{(n)} \right] \). Using the definition of \( \mu_j(n) \) in (2.7), we get

\[
E \left[ T_{(n)}^{(n)} \right] = \sum_{j=1}^{b(n)} j \mu_j(n) = \theta \sum_{j=1}^{b(n)} (x_{n,\alpha})^j = \theta x_{n,\alpha} \frac{(x_{n,\alpha})^{b(n)} - 1}{x_{n,\alpha} - 1} - 1.
\]  

We know from Lemma 2.1 that \( x_{n,\alpha} \to 1 \) and

\[
(x_{n,\alpha})^{b(n)} \sim \left( \frac{n}{\partial \alpha(n)} \log \left( \frac{n}{\partial \alpha(n)} \right) \right)^{b(n)/\alpha(n)}.
\]  

If \( b(n) = o(\alpha(n)/\log(n)) \) then \( (x_{n,\alpha})^{b(n)} \to 1 \) and thus \( E \left[ T_{(n)}^{(n)} \right] \sim b(n) \). However, we can also have \( b(n) \geq c \frac{\alpha(n)}{\log(n)} \) for some \( c > 0 \). Using that \( x_{n,\alpha} - 1 \sim \log(x_{n,\alpha}) \), we immediately obtain

\[
E \left[ T_{(n)}^{(n)} \right] \approx \alpha(n) \log(n) \left( \frac{n}{\partial \alpha(n)} \log \left( \frac{n}{\partial \alpha(n)} \right) \right)^{b(n)/\alpha(n)}.
\]  

This implies that we have for \( n \) large

\[
\frac{\alpha(n)}{\log(n)} \leq E \left[ T_{(n)}^{(n)} \right] \leq \frac{\alpha(n)n^c}{\log(n)},
\]  

where \( c > 0 \) can be chosen arbitrarily. In view of (3.32) and \( b(n) = o(\alpha(n)) \), we use \( \rho = \log^2(n) \) in Lemma 3.7. With this choice of \( \rho \), we immediately get that \( P \left[ T_{(n)}^{(n)} \geq \rho E \left[ T_{(n)}^{(n)} \right] \right] = O(n^{-A}) \) where \( A > 0 \) is arbitrary. Inserting this into (3.26), with \( A = 2 \), we get

\[
d_{b(n)} \leq \max_{r \leq \rho E \left[ T_{(n)}^{(n)} \right]} \left( 1 - \frac{P \left[ T_{(n)}^{(n)} = n - r \right]}{P \left[ T_{(n)}^{(n)} = n \right]} \right) + O(n^{-2}).
\]  

We look next at \( T_{(n)}^{(n)} \) for \( A \). Using that all \( Y_j \) are independent, we get that the probability generating function of \( T_{(n)}^{(n)} \) is

\[
E \left[ z^{T_{(n)}^{(n)}} \right] = \exp \left( \sum_{j=b(n)+1}^{\alpha(n)} \mu_j(n)(z^j - 1) \right).
\]
Using that $\mu_j(n) = \vartheta^{(x_{n,\alpha})_j}$, we get

$$
P \left[ T_{b(n)\alpha(n)}^{(n)} = n - r \right] = \exp \left[ - \sum_{j=b(n)+1}^{\alpha(n)} \mu_j(n) \right] x_{n,\alpha}^{n-r} [z^n] z^n \exp \left( \vartheta \sum_{j=b(n)+1}^{\alpha(n)} \frac{1}{j} z^j \right).$$

Similarly, we obtain

$$
P \left[ T_{0\alpha(n)}^{(n)} = n \right] = \exp \left( - \sum_{j=1}^{\alpha(n)} \mu_j(n) \right) x_{n,\alpha}^{n} [z^n] \exp \left( \vartheta \sum_{j=1}^{b(n)} \frac{1}{j} z^j \right) \exp \left( \vartheta \sum_{j=b(n)+1}^{\alpha(n)} \frac{1}{j} z^j \right).$$

Thus we have to determine for $r \leq \rho \mathbb{E} \left[ T_{b(n)}^{(n)} \right]$ the asymptotic behaviour of

$$
[z^n] z^n \exp \left( \vartheta \sum_{j=b(n)+1}^{\alpha(n)} \frac{1}{j} z^j \right) \quad \text{and} \quad [z^n] \exp \left( \vartheta \sum_{j=1}^{b(n)} \frac{1}{j} z^j \right) \exp \left( \vartheta \sum_{j=b(n)+1}^{\alpha(n)} \frac{1}{j} z^j \right).
$$

We do this with Proposition 3.4. We use for both the triangular array

$$q = (q_{j,n})_{1 \leq j \leq \alpha(n), n \in \mathbb{N}} \text{ with } q_{j,n} = \vartheta^{1_{(b(n)+1 \leq j \leq \alpha(n))}}. \quad (3.35)$$

Furthermore, we use the perturbations $f_{1,n}(z) = z^r$ for the first and $f_{2,n}(z) = \exp \left( \vartheta \sum_{j=1}^{b(n)} \frac{1}{j} z^j \right)$ for the second expression. We thus have to show that $q$ and $f_{1,n}(z)$ and $f_{2,n}(z)$ are admissible, see Definitions 3.2 and 3.3. We now have

**Lemma 3.8.** Let $b = o(\alpha(n))$ and define $x_n$ to be the solution of the equation

$$n = \vartheta \sum_{j=b(n)+1}^{\alpha(n)} x_j^n. \quad (3.36)$$

We then have $x_{n,\alpha} \leq x_n \leq x_{n,\alpha-b(n)}$ and $|x_n - x_{n,\alpha}| = O \left( \frac{1}{\alpha(n)} \right)$. Furthermore the triangular array $q$ in (3.35) is admissible.

**Proof.** We have by definition that $x_{n,\alpha} \leq x_n$. Further, $x_{n,\alpha-b}$ is the solution of

$$n = \sum_{j=1}^{\alpha(n)-b(n)} (x_{n,\alpha-b})^j.$$

Since $\alpha(n) < n$, we have $x_n \geq 1$ and $x_{n,\alpha-b} \geq 1$. This implies that $x_n \leq x_{n,\alpha-b}$. Lemma 2.1 now implies

$$|x_n - x_{n,\alpha}| = x_n - x_{n,\alpha} \leq x_{n,\alpha-b} - x_{n,\alpha} = O \left( \frac{1}{\alpha(n)} \right).$$
Further, \( x_{n,\alpha} \) and \( x_{n,\alpha-b} \) are admissible by Lemma 2.1. Thus \( x_{n,\alpha} \leq x_n \leq x_{n,\alpha-b} \) together with Equation (2.10) immediately shows that \( x_n \) fulfills Condition (1) in Definition 3.2. Furthermore, we also get
\[
\log(x_n) \approx \frac{\log(n)}{\alpha(n)} \quad \text{and} \quad x_n - 1 \approx \frac{\log(n)}{\alpha(n)}.
\] (3.37)

To see that \( x_n \) fulfills Condition (2), one uses (3.37) and the identity
\[
\sum_{j=0}^{d} j q^j = \frac{dq^{d+1}}{q-1} - \frac{q(d-1)}{(q-1)^2} \quad \text{for all} \quad d \in \mathbb{N}, q \neq 0.
\] (3.38)

Condition (3) is obvious. Thus \( q \) is admissible. \( \square \)

We now can show

**Lemma 3.9.** The sequences \((f_{1,n})_{n\in\mathbb{N}}\) with \(f_{1,n} = z^r\) is admissible for all \(r = o\left(n^{\frac{2}{5}}\alpha^{\frac{2}{5}}\right)\).

Further, \((f_{2,n})_{n\in\mathbb{N}}\) with \(f_{2,n} = \exp\left(\sum_{j=1}^{b(n)} \frac{1}{j} z^j\right)\) is admissible.

**Proof.** We start with \((f_{1,n})_{n\in\mathbb{N}}\). Since all \(f_{1,n} = z^r\), the first two conditions of Definition 3.3 are fulfilled with \(\delta = N = 1\) and \(K = 0\) for all \(r\). We now have
\[
\|f_{1,n}\|_n \leq n^{-\frac{2}{5}} (\alpha(n))^{-\frac{2}{5}} r x_n^{-1}.
\]

Since \(x_n \to 1\), we have \(\|f_{1,n}\|_n \to 0\) if and only if \(r = o(n^{\frac{2}{5}})\). This completes the proof of the first half of the statement. For \((f_{2,n})_{n\in\mathbb{N}}\), we also have only to check the third condition. Lemma 2.1 implies that \(x_n - 1 \geq c\log(n)/\alpha(n)\) for some \(c > 0\). Since \(x_{n,\alpha} \leq x_n \leq x_{n,\alpha-b}\) and \(b = o(\alpha(n))\), we get with Lemma 2.1
\[
\frac{|f_{2,n}(z)|}{|f_{2,n}(x_n)|} \leq \sum_{j=0}^{b(n)-1} n_j = \frac{x_{n,\alpha-b} - 1}{x_{n,\alpha} - 1} = O(n^{\alpha(n)}) \quad \text{for all} \quad z \quad \text{with} \quad |z| = x_n,
\]

where \(\epsilon > 0\) can be chosen arbitrarily small. We thus have \(\|f_{2,n}\|_n \leq n^{-\frac{2}{5}+\epsilon} (\alpha(n))^{\frac{2}{5}}\).

Since \(\alpha(n) \leq n^{a_2}\) with \(a_2 < 1\), we see that \(\|f_{2,n}\|_n \to 0\) for \(\epsilon > 0\) small enough. \( \square \)

We know from (3.32) that \(\mathbb{E}\left[T_{0b(n)}^{(n)}\right] \leq \frac{\alpha(n)n^\epsilon}{\log(n)}\) for each \(\epsilon > 0\) and \(n\) large enough. This shows that we can use Proposition 3.4 to compute \(\mathbb{P}\left[T_{0b(n)}^{(n)}(n) = n - r\right] \quad \text{and} \quad \mathbb{P}\left[T_{0a(n)}^{(n)} = n\right]\) for \(r \leq R\mathbb{E}\left[T_{0b(n)}^{(n)}\right]\). We thus have
\[
\mathbb{P}\left[T_{0b(n)a(n)}^{(n)} = n - r\right] = x_{n,\alpha}^{-r} \exp\left(-\sum_{j=1}^{b(n)} \frac{1}{j} (x_n^j - x_{n,\alpha}^j)\right) (1 + R_n),
\] (3.39)
where

\[ R_n = O \left( \frac{\alpha(n)}{n} + \|f_{1,n}\|_n + \|f_{2,n}\|_n \right). \]

Note that the implicit constant in the error term in Proposition 3.4 only depends on the used \( K, N \) and \( \delta \). Since we use for each \( r \leq \rho E \left[ T_{1b(n)}^{(n)} \right] \) the same \( K, N \) and \( \delta \), we get that \( R_n \) is uniform in \( r \). We now have to distinguish the two cases \( b(n) = o(\alpha(n)) \) and \( b(n) = o \left( \frac{\alpha(n)}{\log(n)} \right) \) for the error terms in (2.18) and (2.19). In the case \( b(n) = o(\alpha(n)) \), we get with (3.32) and the proof of Lemma 3.9 that

\[ \|f_{1,n}\|_n \leq \frac{r}{n \Psi(\alpha(n))^{\frac{\alpha(n)}{n}}} \leq \frac{\rho E \left[ T_{1b(n)}^{(n)} \right]}{n \Psi(\alpha(n))^{\frac{\alpha(n)}{n}}} = O \left( n^\epsilon \left( \frac{\alpha(n)}{n} \right)^{\frac{\alpha(n)}{2n}} \right) \]

for each \( \epsilon > 0 \). We thus have that \( R_n \) is as in (2.18). In the case \( b(n) = o \left( \frac{\alpha(n)}{\log(n)} \right) \), we have \( E \left[ T_{1b(n)}^{(n)} \right] \sim b(n) \). Using this, we immediately get that \( R_n \) is as in (2.19).

It thus remains to compute the asymptotic behaviour of the main term in (3.39). We thus need an estimate for \( x_n^{b(n)} - x_n^{b(n)} \). Unfortunately, the bounds obtained from the Lemmas 3.8 and 2.1 are not strong enough. To overcome this issue, let us consider for \( y \in \mathbb{R} \) the equation

\[ y e^{\alpha(n)y} = ny. \tag{3.40} \]

It is straightforward to see that this equation has for \( \frac{n}{\alpha(n)} < e \) two solutions. We denote these by \( y_{n,0,0} \) and \( y_{n,0} \) with \( 0 < y_{n,0,0} < y_{n,0} \). It is straightforward to see that \( y_{n,0,0} \sim \frac{\alpha(n)}{n} \) and \( y_{n,0} \sim \frac{\log(n)/\alpha(n)}{\alpha(n)} \) as \( n \to \infty \). We have

**Lemma 3.10.** We have

\[ \alpha(n) y_{n,\alpha} = \log \left( \frac{n}{\alpha(n)} \log \left( \frac{n}{\alpha(n)} \right) \right) + O \left( \frac{\log \log(n)}{\log(n)} \right). \tag{3.41} \]

Furthermore, we have for \( b = o(\alpha(n)) \) that

\[ \log(x_{n,\alpha}) = y_{n,\alpha} + O \left( \frac{1}{n \log(n)} \right) \quad \text{and} \quad \log(x_n) = y_{n,\alpha} + O \left( \frac{e^{b(n)y_{n,\alpha}}}{n \log(n)} \right). \tag{3.42} \]

We first complete our computations of the main term in (3.39) with Lemma 3.10 and then give the proof of Lemma 3.10. We have

\[ \partial \sum_{j=1}^{b(n)} \left( x_n^j - x_{n,\alpha}^j \right) = \partial \sum_{j=0}^{b(n)-1} \int_{x_{n,\alpha}}^{x_n} v^j dv = \partial \int_{x_{n,\alpha}}^{x_n} \frac{v^{b(n)} - 1}{v - 1} dv \leq \frac{\partial}{x_{n,\alpha} - 1} \int_{x_{n,\alpha}}^{x_n} v^{b(n)} dv \]

\[ \quad = \frac{\partial}{x_{n,\alpha} - 1} \left( \frac{(x_n)^{b(n)+1}}{b(n)+1} - \frac{(x_{n,\alpha})^{b(n)+1}}{b(n)+1} \right). \tag{3.43} \]
We use (3.42) and get for some $\epsilon > 0$
\[(x_n)^{b(n)+1} - (x_{n,\alpha})^{b(n)+1} = (x_{n,\alpha})^{b(n)+1} \left( \exp \left( (b(n) + 1)(\log x_n - \log x_{n,\alpha}) \right) - 1 \right)\]
\[= (x_{n,\alpha})^{b(n)+1} \left( \exp \left( (b(n) + 1)\mathcal{O} \left( \frac{e^{b(n)y_{n,\alpha}}}{n \log n} \right) \right) - 1 \right)\]
\[= (x_{n,\alpha})^{b(n)+1}(b(n) + 1)\mathcal{O} \left( \frac{e^{b(n)y_{n,\alpha}}}{n \log n} \right).\]

Equation 3.41 and Lemma 2.1 imply that $e^{b(n)y_{n,\alpha}} = \mathcal{O} (n^\epsilon)$ and $(x_{n,\alpha})^{b(n)+1} = \mathcal{O} (n^\epsilon)$, where $\epsilon > 0$ can be chosen arbitrarily small. Using this and (3.37), we get
\[\vartheta \sum_{j=1}^{b(n)} \frac{1}{j} (x_n^j - x_{n,\alpha}^j) = \mathcal{O} \left( \frac{(b(n) + 1)e^{b(n)y_{n,\alpha}}(x_{n,\alpha})^{b(n)+1}}{n \log^2(n)} \right) = \mathcal{O} \left( \frac{b(n) + 1}{n^{1-2\epsilon} \log^2(n)} \right).\]

Inserting this into (3.43) gives
\[\frac{\mathbb{P}[T_{b(n)\alpha(n)} = n - r]}{\mathbb{P}[T_{0\alpha(n)} = n]} \geq \exp \left( \mathcal{O} \left( \frac{b(n) + 1}{n^{1-2\epsilon} \log^2(n)} \right) \right) (1 + R_n)\]
\[= 1 + \mathcal{O} \left( n^\epsilon \left( \frac{\alpha(n)}{n} \right)^{\frac{1}{2\epsilon}} \right).\]

This equation together with (3.33) completes the proof of Theorem 2.8.

Proof of Lemma 3.10. We start with (3.41). We insert the approach
\[y = \frac{1}{\alpha(n)} \log \left( \frac{n}{\vartheta \alpha(n)} \log \left( \frac{n}{\vartheta \alpha(n)} \right) \right) + v\]
with $v \in \mathbb{R}$ into (3.40). This leads to the equation
\[\log \left( \frac{n}{\vartheta \alpha(n)} \right) e^{\alpha(n)v} = \log \left( \frac{n}{\vartheta \alpha(n)} \log \left( \frac{n}{\vartheta \alpha(n)} \right) \right) + \alpha(n)v.\] (3.44)

Note that we have
\[\log(y) \leq \log(y \log(y)) \leq (1 + \epsilon) \log(y)\] (3.45)
for all $\epsilon > 0$ and $y$ large enough. Using this, it is straightforward to see that equation (3.44) has exactly one solution in the region $v \geq 0$ and that this solution has to be $o \left( \frac{1}{\alpha(n)} \right)$ as $n \to \infty$. To obtain a lower bound for $v$, we use the inequality $e^x \leq 1 + 2x$ for $0 \leq x \leq \log 2$. Thus $v$ is larger than the solution $v'$ of the equation
\[\log \left( \frac{n}{\vartheta \alpha(n)} \right) (1 + 2\alpha(n)v') = \log \left( \frac{n}{\vartheta \alpha(n)} \log \left( \frac{n}{\vartheta \alpha(n)} \right) \right) + \alpha(n)v'.\] (3.46)
A simple computation gives

\[ v' = \frac{\log \log \left( \frac{n}{\vartheta \alpha(n)} \right)}{2\alpha(n) \log \left( \frac{n}{\vartheta \alpha(n)} \right) + \alpha(n)} \]  

(3.47)

This establishes a lower bound for \( v \). For an upper bound, we argue similarly with \( 1 + x \leq e^x \) for \( x \geq 0 \). This completes the proof of (3.41).

We prove (3.42) only for \( x_n \). The asymptotics for \( x_{n,\alpha} \) then follows immediately by inserting \( b = 0 \) into the asymptotics for \( x_n \). The defining equation (3.36) of \( x_n \) has exactly one solution can be rewritten as

\[ \vartheta(x_n)^{\alpha(n)} - \vartheta(x_n)^{b(n)} = n \left( 1 - (x_n)^{-1} \right) \]

(3.48)

We now insert \( x_n = e^y \). This gives

\[ \vartheta e^{\alpha(n)y} - \vartheta e^{b(n)y} = n \left( 1 - e^{-y} \right) \]

(3.49)

The equation (3.49) has exactly one solution in the region \( y > 0 \). Further, both sides of (3.49) are monotone increasing functions of \( y \). Inserting \( y = y_{n,\alpha} \pm \frac{c}{\alpha(n)} \) with \( c > 0 \) into (3.49) and using (3.41) shows that the RHS of (3.49) behaves like

\[ n \left( 1 - e^{-y_{n,\alpha} \pm \frac{c}{\alpha(n)}} \right) \sim \frac{n}{\alpha(n)} \log \left( \frac{n}{\vartheta \alpha(n)} \right) \]

(3.50)

On the other hand, the LHS of (3.49) behaves like

\[ \vartheta e^{\alpha(n)(y_{n,\alpha} \pm \frac{c}{\alpha(n)})} - \vartheta e^{b(n)(y_{n,\alpha} \pm \frac{c}{\alpha(n)})} \sim e^{\pm c} \frac{n}{\alpha(n)} \log \left( \frac{n}{\vartheta \alpha(n)} \right) \]

(3.51)

Using (3.45), we immediately see that the solution of (3.49) has to be in the interval \( [y_{n,\alpha} - \frac{c}{\alpha(n)}, y_{n,\alpha} + \frac{c}{\alpha(n)}] \). We now use the approach \( y = y_{n,\alpha} + v \). Clearly, we must have \( v = o \left( \frac{1}{\alpha(n)} \right) \). We now argue as for (3.41). To get a lower bound for \( v \), we use \( 1 + x \leq e^x \) and \( 1 - e^{-x} \leq x \). This leads to the equation

\[ \vartheta e^{\alpha(n)(y_{n,\alpha} + (1 + \alpha(n))v)} - \frac{3}{2} \vartheta e^{b(n)(y_{n,\alpha})} = n(y_{n,\alpha} + v') \]

Using the definition of \( y_{n,\alpha} \) in (3.40), we immediately get

\[ v' = \frac{3\vartheta^{b(n)(y_{n,\alpha})}}{2\vartheta e^{\alpha(n)y_{n,\alpha} \alpha(n)} - 2n} = \frac{3\vartheta^{b(n)(y_{n,\alpha})}}{2ny_{n,\alpha} \alpha(n) - 2n} \sim \frac{3\vartheta^{b(n)(y_{n,\alpha})}}{2n \log(n)} \]

The upper bound is obtained similarly. This completes the proof.
3.6. Proof of Theorem 2.9

We give here the proof for the case $K = 1$. The reason is that the arguments for the cases $K > 1$ and $K = 1$ are almost identical, but unfortunately the notation for the case $K > 1$ is more cumbersome. We thus write $m(n)$ and $\mu_{m(n)}$ instead of $m_1(n)$ and $\mu_{m_1(n)}(n)$. This mainly simplifies the notation, but does not change the argument used. As in [4], the proof will be based upon point-wise convergence of moment-generating functions. Replacing $s$ by $s/\sqrt{\mu_{m(n)}}$ in (3.53), we get

$$M_n(s) := \mathbb{E}_{n,\alpha} \left[ \exp \left( \frac{s}{\sqrt{\mu_{m(n)}}} C_{m(n)} \right) \right]$$

(3.52)

$$= \frac{1}{Z_{n,\alpha}} [z^n] \exp \left( \vartheta \sqrt{\mu_{m(n)}} \frac{z^{m(n)}}{m(n)} + \vartheta \sum_{1 \leq j \leq \alpha(n), j \neq m} z^j \right).$$

(3.53)

In order to determine the asymptotic behaviour of $M_n(s)$, we apply Proposition 3.4 with the triangular array $q = (q_{j,n})_{1 \leq j \leq \alpha(n), n \in \mathbb{N}}$ with

$$q_{j,n} = \begin{cases} 
0 & \text{if } j > \alpha(n) \\
\vartheta \exp \left( \frac{s}{\sqrt{\mu_{m(n)}}} \right) & \text{if } j = m(n) \\
\vartheta & \text{otherwise},
\end{cases}$$

(3.54)

together with $f_n(z) = 1$ for all $n$. We thus have to show that $q$ and the sequence $(f_n)_{n \in \mathbb{N}}$ are both admissible, see Definition 3.2 and 3.3. The sequence $(f_n)_{n \in \mathbb{N}}$ is admissible for all triangular arrays. Thus we have only to show that $q$ is admissible. Hence, we have to study the solution $x_{n,q}$ of the equation (3.5). Since this solution depends on the parameter $s$, we write $x_{n,q}(s)$ instead of $x_{n,q}$. Also, we will write $\lambda_{2,n}(s)$ for $\lambda_{2,n,\alpha,q}$ with $\lambda_{2,n,\alpha,q}$ as in (3.6). We now show

**Lemma 3.11.** Let $q$ be as in (3.54) and $x_{n,q}(s)$ be defined as in (3.5). Suppose that $\mu_{m(n)} \to \infty$ with $\mu_{m(n)}$ as in (2.7). Then we have, locally uniformly in $s \in \mathbb{R}$, that

$$\alpha(n) \log (x_{n,q}(s)) \sim \log \left( \frac{n}{\vartheta \alpha(n)} \right) \log \left( \frac{n}{\vartheta \alpha(n)} \right).$$

(3.55)

In particular, if $n$ is large enough,

$$x_{n,q}(s) \geq 1 \text{ and } \lim_{n \to \infty} x_{n,q}(s) = 1.$$ 

Furthermore, we have

$$\lambda_{2,n}(s) \sim n \alpha(n)$$

(3.56)

locally uniformly in $s$ with $\lambda_{2,n}(s) = \lambda_{2,n,\alpha,q}$ as in (3.6).
Proof. We use Lemma 2.1 to prove Lemma 3.11. Recall that \( x_{n,\alpha}(c) \) is defined in (2.9) for \( c > 0 \) as the solution of
\[
cn = \vartheta \sum_{j=1}^{\alpha(n)} (x_{n,\alpha}(c))^j.
\]
Furthermore \( x_{n,q}(s) \) is the solution of the equation
\[
n = \vartheta \left( e^{\sqrt{\mu_m(n)}} - 1 \right) (x_{n,q}(s))^{m(n)} + \vartheta \sum_{j=1}^{\alpha(n)} (x_{n,q}(s))^j.
\] (3.57)
We now assume that \( 0 \leq s \leq U \) with \( U > 0 \) an arbitrary, but fixed, real number. Since \( e^{\sqrt{\mu_m(n)}} \geq 1 \), we get
\[
x_{n,q}(s) \leq x_{n,\alpha}(1) = x_{n,\alpha},
\] (3.58)
where \( x_{n,\alpha} \) is as in (2.6). Using the definition of \( \mu_m(n) \) together with \( s \leq U \) and \( \mu_m(n) \to \infty \), we obtain for \( n \) large
\[
\left( e^{\sqrt{\mu_m(n)}} - 1 \right) (x_{n,q}(s))^{m(n)} \leq \frac{2U}{\vartheta \sqrt{\mu_m(n)}} (x_{n,\alpha})^{m(n)} = \frac{2U}{\vartheta} \sqrt{\mu_m(n)} (x_{n,\alpha})^{m(n)}
\]
\[
\leq \frac{2U}{\vartheta} \sqrt{\alpha(n)} (x_{n,\alpha})^{\alpha(n)}.
\]
Applying Lemma 2.1 for \( x_{n,\alpha} = x_{n,\alpha}(1) \), we get for \( n \) large
\[
\left( e^{\sqrt{\mu_m(n)}} - 1 \right) (x_{n,q}(s))^{m(n)} \leq \frac{4U}{\vartheta} \sqrt{n \log \left( \frac{n}{\vartheta \alpha(n)} \right)} \leq n^{1/2+\epsilon},
\] (3.59)
for \( \epsilon > 0 \) small. Inserting this into (3.57), we get
\[
\vartheta \sum_{j=1}^{\alpha(n)} (x_{n,q}(s))^j = n - \vartheta \left( e^{\sqrt{\mu_m(n)}} - 1 \right) (x_{n,q}(s))^{m(n)} \geq n(1 - n^{-1/2+\epsilon}).
\] (3.60)
Using the definition if \( x_{n,\alpha}(c) \), we see that
\[
x_{n,\alpha} \left( 1 - n^{-1/2+\epsilon} \right) \leq x_{n,q}(s) \leq x_{n,\alpha}(1).
\] (3.60)
Applying Lemma 2.1 to \( x_{n,\alpha} \left( 1 - n^{-1/2+\epsilon} \right) \) and \( x_{n,\alpha}(1) \) immediately completes the proof for \( 0 \leq s \leq U \). The argumentation for \( -U \leq s \leq 0 \) is similar and we thus omit it. □

Lemma 3.11 implies that \( x_{n,q}(s) \) with \( q \) in (3.54) is admissible. Thus we can apply Proposition 3.4. We obtain for each \( s \geq 0 \) that
\[
M_n(s) = \frac{1}{Z_{n,\alpha}} \frac{\exp(h_n(s))}{\sqrt{2\pi \alpha_n(s)}} (1 + o(1)),
\] (3.61)
where
\[
    h_n(s) = \vartheta \left( e^{\sqrt{\mu_m(n)}} - 1 \right) \frac{(x_{n,q}(s))^{m(n)}}{m(n)} + \sum_{j=1}^{\alpha(n)} \frac{(x_{n,q}(s))^j}{j} - n \log(x_{n,q}(s)).
\] (3.62)

Since \( M_n(0) = 1 \), we have
\[
    \frac{1}{Z_{n,\alpha}} \exp(h_n(0)) \sqrt{2\pi \lambda_{2,n}(0)} \to 1 \quad \text{as} \quad n \to \infty.
\]

Our aim is to use this result to complete the proof of Theorem 2.9. We observe from (3.56) that the leading coefficient of \( \lambda_{2,n}(s) \) is independent of \( s \). Therefore, we have proven Theorem 2.9 if we can show that for each \( s \geq 0 \)
\[
    h_n(s) = h_n(0) + s\sqrt{\mu_m(n)} + \frac{s^2}{2} + o(1) \quad \text{as} \quad n \to \infty.
\] (3.63)

We begin with the derivatives of \( x_{n,q}(s) \)

**Lemma 3.12.** The function \( s \mapsto x_{n,q}(s) \) is for each \( n \) infinitely often differentiable.

Further, we have
\[
    \frac{x'_{n,q}(s)}{x_{n,q}(s)} = -\exp\left( \frac{s}{\sqrt{\mu_m(n)}} \right) \frac{(x_{n,q}(s))^{m(n)}}{\sqrt{\mu_m(n)} \lambda_{2,n}(s)}.
\] (3.64)

**Proof.** Let \( n \) be fixed. Since all coefficients of \( q \) in (3.54) are non-negative and not all 0, it follows that the equation (3.5) has for each \( s \geq 0 \) exactly one solution. Thus the function \( s \mapsto x_{n,q}(s) \) is a well defined function on \( [0, \infty) \). Applying the implicit function theorem to the function
\[
    g(s, x) = \vartheta \left( e^{\sqrt{\mu_m(n)}} - 1 \right) x^{m(n)} + \vartheta \sum_{j=1}^{\alpha(n)} x^j
\]
and using that \( \frac{\partial}{\partial x} g(s, x) > 0 \) for \( x > 0 \) completes the proof. \( \Box \)

Applying Lemma 3.12 to \( h_n(s) \), we obtain

**Lemma 3.13.** We have
\[
    h'_n(s) = \vartheta \frac{e^{\sqrt{\mu_m(n)}}}{\sqrt{\mu_m(n)}} \frac{(x_{n,q}(s))^{m(n)}}{m(n)};
\] (3.65)
\[
    h''_n(s) = \frac{1}{\sqrt{\mu_m(n)}} h'_n(s) - \frac{(m(n))^2}{\lambda_{2,n}(s)} (h'_n(s))^2;
\] (3.66)
\[
    h'''_n(s) = \frac{1}{\sqrt{\mu_m(n)}} h''_n(s) - \frac{2(m(n))^2}{\lambda_{2,n}(s)} h'_n(s) h''_n(s) + \frac{\lambda_{3,n}(s) x_{n,q}(s)}{(\lambda_{2,n}(s))^2} (h'_n(s))^2.
\] (3.67)
**Proof.** Equation (3.65) follows immediately from (3.62) and the definition of \( x_{n,q}(s) \). Equation (3.66) and (3.67) follow from Lemma 3.12, equation (3.65) and the definition of \( \lambda_{2,n}(s) \), see Definition 3.2.

Equation (3.63) now follows by using \( x_{n,q}(0) = x_{n,\alpha} \), \( \mu_m = \vartheta \frac{x_m}{m} \), and Lemma 3.11. This completes the proof of Theorem 2.9.

**References**


[18] Benjamin Lees and Lorenzo Taggi. Site monotonicity and uniform positivity for interacting random walks and the spin o(n) model with arbitrary n. Communications in Mathematical Physics, 2019.


