CROSS-MULTIPLICATIVE COALESCENCE AND MINIMAL SPANNING TREES OF IRREGULAR GRAPHS

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Abstract. This paper concentrates on bridging the gap between the Smoluchowski coagulation equations for Marcus-Lushnikov processes and the theory of random graphs with the goal of finding the limiting mean length of a minimal spanning tree on an irregular graph. Specifically, let \( \alpha, \beta > 0 \) and \( L_n = L_n(\alpha, \beta) \) be the length of a minimal spanning tree on a complete bipartite graph \( K_{\alpha[n], \beta[n]} \) with partitions of size \( \alpha[n] = \alpha n + o(\sqrt{n}) \) and \( \beta[n] = \beta n + o(\sqrt{n}) \), and independent uniform edge weights over \([0, 1] \).

There, the following expression for the limiting mean length of the minimal spanning tree is derived via the hydrodynamic limit of a novel cross-multiplicative coalescent process that we introduce and study in this paper:

\[
\lim_{n \to \infty} E[L_n] = \gamma + \frac{1}{\gamma} \sum_{i_1 \geq 1; i_2 \geq 1} \frac{(i_1 + i_2 - 1)!}{i_1! i_2!} \frac{\gamma^{i_1} (i_1 - 1)! i_1^{-1} (i_2 - 1)! i_2^{-1}}{(i_1 + \gamma i_2)^{i_1 + i_2}}
\]

where \( \gamma = \frac{\alpha}{\beta} \). This is a completely new formula for the case of an irregular bipartite graph \( \gamma \neq 1 \). In the case of \( \gamma = 1 \), the above series adds up to

\[
\lim_{n \to \infty} E[L_n] = 2\zeta(3)
\]

as derived in Frieze and McDiarmid [15] for a regular bipartite graph. A generalization of the approach is considered in the discussion section.

1. Introduction

We begin with the following quote from Aldous [1]: *It turns out that there is a large scientific literature relevant to the Marcus-Lushnikov process, mostly focusing on its deterministic approximation. Curiously, this literature has been largely ignored by random graph theorists.* The broader goal of this paper is in bridging the gap between the theory of the Smoluchowski coagulation equations for the Marcus-Lushnikov processes and the random graph theory. This paper attempts to extend the connection between coalescent processes and random graph processes, e.g. Erdős-Rényi random graph evolution as described in Subsection 2.1. In particular, deriving a formula for the limiting length of the minimal spanning tree in a random graph process in terms of the solutions of the Smoluchowski coagulation equations for the corresponding coalescent process.

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In this paper we will concentrate on analyzing the length of the minimal spanning tree as the prime example that demonstrates the usefulness of the Marcus-Lushnikov processes and the coalescence theory in general for answering questions about random graphs. We recall that the asymptotic limit for the mean length of a minimal spanning tree on $K_n$ with independent uniform edge weights over $[0,1]$ was derived in Frieze [14],
\[
\lim_{n \to \infty} E[L_n] = \zeta(3) = \sum_{k=1}^{\infty} \frac{1}{k^3}.
\]
The mean length of a minimal spanning tree on the complete bipartite graph $K_{n,n}$ with independent uniform edge weights over $[0,1]$ was shown in [15] to be
\[
\lim_{n \to \infty} E[L_n] = 2\zeta(3).
\]
In Beveridge et al [5], the minimal spanning tree problem was addressed for $d$-regular graphs. In this paper we will find the mean length of the minimal spanning tree in the case of an irregular complete bipartite graph by means of developing a connection between the coalescence theory and the random graph theory.

We observe that in many cases the cluster dynamics of a random graph process can be replicated with the corresponding coalescent process. For example, the Erdős-Rényi random graph process on $K_n$ can be tied to the $n$-particle multiplicative coalescent. The connection lies in that the probability of two components merging, at a given time, depends only on the number of edges that connect those two components (rather than other structural properties). There are many other, more elaborate examples. The cluster dynamics of a coalescent process (without merger history) is reflected in an auxiliary process called the Marcus-Lushnikov process. The merger dynamics of such coalescent processes corresponds to a greedy algorithm for finding the minimal spanning tree in the respective random graph process. This observation allows us to express the limiting mean length of a minimal spanning tree in terms of the solutions of the Smoluchowski coagulation equations that represent the hydrodynamic limit of the Marcus-Lushnikov process corresponding to the random graph process. As a particular application of the proposed general approach, we find the asymptotic for the mean length of a minimal spanning tree for the complete bipartite graph with partitions of size $\alpha[n] = \alpha n + o(\sqrt{n})$ and $\beta[n] = \beta n + o(\sqrt{n})$.

The paper is organized as follows. In Section 2, the cluster dynamics of the Erdős-Rényi random graph process on $K_n$ is considered. The following theorem is proved using only the weak convergence results that appear in Section 4 of this paper.

**Theorem 1.1 (Theorem 2.1).** Let $L_n$ denote the length of the minimal spanning tree in $K_n$, where edge weights are independent and uniform random variables on $[0,1]$. Then
\[
\lim_{n \to \infty} E[L_n] = \sum_{k=1}^{\infty} \int_{0}^{\infty} \zeta_k(t) dt,
\]
where $\zeta_k(t)$ are the solutions (6) of the corresponding system of reduced Smoluchowski coagulation equations (5).

Although, the above result is not novel [14], its proof given in Subsection 2.4 relies entirely on the use of the hydrodynamic limits from Section 4. At the end of Subsection 2.3 a
general approach for finding \( \lim_{n \to \infty} E[L_n] \) via the Smoluchowski coagulation equations is proposed.

Our main result, proven in Section 3, finds the asymptotics for the mean length of a minimal spanning tree on the complete bipartite graph with partitions of size \( \alpha n \) and \( \beta n \). There too, the probability of two components merging, at a given time, depends only on the number of edges that connect those two components. If connected component \( C_i \) and \( C_j \) have partition sizes \((i_1, i_2)\) and \((j_1, j_2)\) respectively, then there are \(i_1 j_2 + i_2 j_1\) edges which, when opened, would connect \( C_i \) and \( C_j \).

**Theorem 1.2** (Theorem 3.2). Let \( \alpha, \beta > 0 \) and \( L_n = L_n(\alpha, \beta) \) be the length of a minimal spanning tree on a complete bipartite graph \( K_{\alpha[n], \beta[n]} \) with partitions of size

\[
\alpha[n] = \alpha n + o(\sqrt{n}) \quad \text{and} \quad \beta[n] = \beta n + o(\sqrt{n})
\]

and independent uniform edge weights over \([0, 1]\). Then

\[
\lim_{n \to \infty} E[L_n] = \sum_{i_1, i_2} \int_0^\infty \zeta_{i_1, i_2}(t) d(t),
\]

where \( \zeta_{i_1, i_2}(t) \) indexed by \( \mathbb{Z}_+^2 \setminus \{(0, 0)\} \) is the solution of the following system of equations

\[
\frac{d}{dt} \zeta_{i_1, i_2}(t) = - (\beta i_1 + \alpha i_2) \zeta_{i_1, i_2}(t) + \frac{1}{2} \sum_{\ell_1, k_1: \ell_1 + k_1 = i_1, \ell_2, k_2: \ell_2 + k_2 = i_2} (\ell_1 k_2 + \ell_2 k_1) \zeta_{\ell_1, k_2}(t) \zeta_{\ell_2, k_1}(t)
\]

with the initial conditions \( \zeta_{i_1, i_2}(0) = \alpha \delta_{i_1, i_2} \delta_{0, i_2} + \beta \delta_{0, i_1} \delta_{i_1, i_2} \).

The above system of equations is the reduced Smoluchowski coagulation system (21) of a novel cross-multiplicative coalescent process with kernel \( K((i_1, i_2), (j_1, j_2)) = i_1 j_2 + i_2 j_1 \) introduced in Section 3 of this paper. In Section 4, functions \( \zeta_{i_1, i_2}(t) \) are determined to be the hydrodynamic limit for the cross-multiplicative coalescent process via the weak convergence results of Kurtz [10, 20].

Solving the system of equations in Theorem 3.2 we obtain the main result of this paper.

**Theorem 1.3** (Theorem 3.3). Let \( \alpha, \beta > 0, \gamma = \alpha / \beta \), and \( L_n = L_n(\alpha, \beta) \) be the length of a minimal spanning tree on a complete bipartite graph \( K_{\alpha[n], \beta[n]} \) with partitions of size

\[
\alpha[n] = \alpha n + o(\sqrt{n}) \quad \text{and} \quad \beta[n] = \beta n + o(\sqrt{n})
\]

and independent uniform edge weights over \([0, 1]\). Then the limiting mean length of the minimal spanning tree is

\[
\lim_{n \to \infty} E[L_n] = \gamma + \frac{1}{\gamma} \sum_{i_1 \geq 1, i_2 \geq 1} \frac{(i_1 + i_2 - 1)!}{i_1! i_2!} \frac{\gamma^{i_1} i_1^{i_2 - 1} i_2^{i_2 - 1}}{(i_1 + \gamma i_2)^{i_1 + i_2}}.
\]
The above result is novel for $\alpha \neq \beta$, when the complete bipartite graph $K_{\alpha[n],\beta[n]}$ is an irregular graph. For $\alpha = \beta$, Theorem 3.3 recovers the result of Frieze and McDiarmid [15], as stated in the following corollary that we also prove in Section 3.

**Corollary 1.4** (Corollary 3.4). If $\alpha = \beta$, then

$$
\lim_{n \to \infty} E[L_n] = 2\zeta(3).
$$

Finally, in Section 4, the weak convergence results of Kurtz [10, 20] are applied to Marcus-Lushnikov processes with multiplicative and cross-multiplicative kernels. The paper concludes with a discussion in Section 5.

2. Erdős-Rényi process on $K_n$ and multiplicative coalescent

In this section we demonstrate how the weak convergence results of Section 4 can be used for finding the length of the minimal spanning tree in the Erdős-Rényi random graph process on $K_n$. The main result of this section Theorem 2.1 is proved in Subsection 2.4.

2.1. Erdős-Rényi random graph. Recall that Erdős-Rényi random graph is a model on a complete graph of $n$ vertices, $K_n$, where each edge $e$ of $\binom{n}{2}$ edges there is an associated uniform random variable $U_e$ over $[0, 1]$. The random variables $\{U_e\}_e$ are assumed to be independent. For the “time” parameter $p \in [0, 1]$, an edge $e$ is considered “open” if $U_e \leq p$. Erdős-Rényi random graph $G(n, p)$ will consist of all $n$ vertices and all open edges at time $p$. The number of open edges is a binomial random variable with parameters $\binom{n}{2}$ and $p$, and mean value $\binom{n}{2}p \sim np^2$. As we increase $p$, more and more edges open up, new clusters are created, and cluster merges occur. Thus Erdős-Rényi random graph model can be viewed as a dynamical model that describes an evolution of a random graph [8].

If we condition on the number of edges in $G(n, p)$, the graph structure will no longer depend on $p$. Let $\xi_{n,N}$ be the number of components in an Erdős-Rényi random graph with $n$ vertices and $N$ edges. For $t > 0$, letting $N \sim \frac{nt}{2}$, Theorem 6 in [8] by P. Erdős and A. Rényi states that

$$
E[\xi_{n,N}] = \frac{1}{t} \sum_{k=1}^{\infty} \frac{k^{k-2}(te^{-t})^k}{k!} + R_t,
$$

where the error term is

$$
R_t = \begin{cases} 
O\left(\frac{1}{n}\right) & \text{if } 0 < t < 1 \\
O\left(\frac{\log n}{n}\right) & \text{if } t = 1 \\
o(1) & \text{if } t > 1 
\end{cases}.
$$

There $\varphi(t) = \frac{1}{t} \sum_{k=1}^{\infty} \frac{k^{k-2}(te^{-t})^k}{k!}$ reaches its maximum at $t = 1$, and $\varphi(1) = \sum_{k=1}^{\infty} \frac{k^{k-2}e^{-k}}{k!} = \frac{1}{2}$.

Let for $t > 0$,

$$
x(t) := \min\{x > 0 : xe^{-x} = te^{-t}\}. 
$$
In other words \( x(t) \) is the unique \( x \in (0, 1] \) such that \( xe^{-x} = te^{-t} \). Obviously, \( x(t) = t \) for \( 0 < t \leq 1 \). It was pointed out by P. Erdős and A. Rényi that \( \varphi(t) \) in the equation (2) can be represented via \( x(t) \) as follows,

\[
\varphi(t) = \frac{x(t) - x^2(t)}{2t}.
\]

Observe that here, since we are letting \( N \sim \frac{tn}{2} \), parameter \( t \) is essentially equivalent to \( np \). So \( t \) is a scaled time parameter.

### 2.2. Multiplicative Coalescent.

A general finite coalescent process begins with \( n \) singletons (clusters of mass one). The cluster formation is governed by a symmetric collision rate kernel \( K(i, j) = K(j, i) > 0 \). Specifically, a pair of clusters with masses (weights) \( i \) and \( j \) coalesces at the rate \( K(i, j)/n \), independently of the other pairs, to form a new cluster of mass \( i + j \). The process continues until there is a single cluster of mass \( n \). See [26, 2, 4, 3, 11] and references therein.

Formally, for a given \( n \) consider the space \( \mathcal{P}_n \) of partitions of \( [n] = \{1, 2, \ldots, n\} \). Let \( \Pi_0(n) \) be the initial partition in singletons, and \( \Pi_t(n) (t \geq 0) \) be a strong Markov process such that \( \Pi_t(n) \) transitions from partition \( \pi \in \mathcal{P}_n \) to \( \pi' \in \mathcal{P}_n \) with rate \( K(i, j)/n \) provided that partition \( \pi' \) is obtained from partition \( \pi \) by merging two clusters of \( \pi \) of weights \( i \) and \( j \). If \( K(i, j) \equiv 1 \) for all positive integer masses \( i \) and \( j \), the process \( \Pi_t(n) \) is known as Kingman’s \( n \)-coalescent process. If \( K(i, j) = i + j \) the process is called \( n \)-particle additive coalescent. Finally, if \( K(i, j) = ij \) the process is called \( n \)-particle multiplicative coalescent. The so called Marcus-Lushnikov process

\[
ML_n(t) = \left( \zeta_{1,n}(t), \zeta_{2,n}(t), \ldots, \zeta_{n,n}(t), 0, 0, \ldots \right)
\]

is an auxiliary process to the corresponding coalescent process that keeps track of the numbers of clusters in each weight category. Here \( \zeta_{k,n}(t) \) denotes the number of clusters of weight \( k \) at time \( t \geq 0 \). See [22] and [21] for the original papers by Marcus and Lushnikov. The latter work considered the gelation phenomenon emerging in some of the Marcus-Lushnikov processes. The Marcus-Lushnikov process does not differentiate between the clusters of the same weight, and therefore does not keep track of the merger history of the \( n \)-particle coalescent process.

Let the number of vertices in a connected component of a random graph be referred to as a weight of the cluster (or cluster size). Consider the Marcus-Lushnikov process \( ML_n(t) \) corresponding to the multiplicative coalescent process of \( n \) particles. Since the coalescent process begins with \( n \) singletons, \( ML_n(0) = (n, 0, 0, \ldots) \). By construction, the process \( ML_n(t) \) describes cluster size dynamics of the Erdős-Rényi random graph process \( G(n, p) \) with \( p = 1 - e^{-t/n} \). Here the scaled time parameter in the Erdős-Rényi process is \( np = n(1 - e^{-t/n}) \sim t \). Thus the time scale is consistent with the one used in [8] by P. Erdős and A. Rényi. This Marcus-Lushnikov process keeps track of the history of cluster mergers and cluster sizes, but not of individual clusters’ history. Let \( \zeta_{k,n}(t) \) be the
number of clusters of mass $k$ in a multiplicative coalescent process of $n$ particles at time $t$. The deterministic dynamics of the limiting fractions $\zeta_k(t) = \lim_{n \to \infty} \frac{\zeta_{k,n}(t)}{n}$ is described by the Smoluchowski coagulation equations [2, 25, 28] as follows

$$\frac{d}{dt} \zeta_k = -k\zeta_k \sum_{j=1}^{\infty} j \zeta_j + \frac{1}{2} \sum_{j=1}^{k-1} j(k-j)\zeta_j \zeta_{k-j} \quad (k = 1, 2, \ldots) \hspace{1em} \text{with} \hspace{1em} \zeta_k(0) = \delta_{1,k}. \quad (4)$$

The dynamics of the total mass $\sum_{j=1}^{\infty} j \zeta_j$ begins with $\sum_{j=1}^{\infty} j \zeta_j(0) = 1$, and following McLeod [23], we have

$$\frac{d}{dt} \sum_{j=1}^{\infty} j \zeta_j = \sum_{j=1}^{\infty} j \frac{d}{dt} \zeta_j = -\sum_{i,j=1}^{\infty} ij^2 \zeta_i \zeta_j + \frac{1}{2} \sum_{i,j=1}^{\infty} \sum_{i=1}^{j-1} (i + (j - i))i(j - i)\zeta_i \zeta_{j-i}$$

$$= -\sum_{i,j=1}^{\infty} ij^2 \zeta_i \zeta_j + \frac{1}{2} \sum_{i,j=1}^{\infty} (i + j)ij \zeta_i \zeta_j = 0$$

provided convergence of $\sum_{j=1}^{\infty} j^2 \zeta_j(t)$. Thus there exists a time $T_{gel} \in [0, \infty]$, defined as

$$T_{gel} := \sup \{ t > 0 : \sum_{j=1}^{\infty} j^2 \zeta_j(t) \text{ converges} \},$$

such that the following conservation of mass formula is satisfied up to time $T_{gel}$,

$$\sum_{j=1}^{\infty} j \zeta_j(t) = 1.$$

Time $T_{gel} > 0$, if finite, is called the gelation time. The kernel function $K(\cdot, \cdot)$ for which such $T_{gel} < \infty$ is called the gelling kernel. A more general discussion on gelation is given in Subsection 2.5. It is well known [23] that the multiplicative kernel $K(i, j) = ij$ is a gelation kernel with the gelling time $T_{gel} = 1$. Indeed, as it was done in [23], the Smoluchowski coagulation equations reduce to

$$\frac{d}{dt} \zeta_k = -k\zeta_k + \frac{1}{2} \sum_{j=1}^{k-1} j(k-j)\zeta_j \zeta_{k-j} \quad (k = 1, 2, \ldots) \hspace{1em} \text{with} \hspace{1em} \zeta_k(0) = \delta_{1,k} \quad (5)$$

which can be solved explicitly:

$$\zeta_k(t) = \frac{k^{k-2}t^{k-1}}{k!} e^{-kt} \quad \text{for} \quad t \geq 0. \quad (6)$$

The above system of equations (5) will be called the reduced Smoluchowski system. The reason for reducing the system is that the decay rate of $k\zeta_k \sum_{j=1}^{\infty} j \zeta_j$ in (4) was representing the gravitation of clusters of size $k$ towards all the clusters that are accounted for in the system (4). The problem is that a cluster of an exceptionally large size, say $\epsilon n$, in a single
quantity will not be accounted for in (4). Yet, such large cluster has to contribute \( \epsilon k \zeta_k \) to the decay rate. Replacing the decay rate with \( k \zeta_k \) in (5) resolves this issue as the new rate accounts for the gravitation of a cluster of a given size \( k \) towards all clusters in the Marcus-Lushnikov process, whose weights add up to \( n - k = n(1 + O(n^{-1})) \).

The system (5) is also known as the Flory’s coagulation system (named after Flory [13]). Notice, that for the solution (6) of system (5),

\[
\begin{cases}
\sum_{k=1}^{\infty} k \zeta_k(t) = 1 & \text{if } t \leq T_{gel} \\
\sum_{k=1}^{\infty} k \zeta_k(t) < 1 & \text{if } t > T_{gel}.
\end{cases}
\]

The phenomenon of losing total mass after a certain finite time \( T_{gel} \) is called gelation. It is an important phenomenon that was studied extensively in the coagulations equations literature. See [1, 2, 30, 18, 21].

The hydrodynamic limit \( \lim_{n \to \infty} \zeta_{k,n}(t) \) is proven in formula (42) of Subsection 4.2 of this paper, where \( \zeta_k(t) \) is the solution (6) of the reduced Smoluchowski system (5).

Recall the function \( x(t) \) defined in (3). It was observed in [23] that since \( \sum_{k=1}^{\infty} \frac{k^{k-1}(te^{-t})^k}{k!} = x(t), \) the first moment of \( \zeta_k, \)

\[
\sum_{k=1}^{\infty} k \zeta_k(t) = \frac{1}{t} \sum_{k=1}^{\infty} \frac{k^{k-1}(te^{-t})^k}{k!} = \frac{x(t)}{t}.
\]

Thus, \( \sum_{k=1}^{\infty} k \zeta_k(t) = 1 \) if and only if \( t \leq 1 \), and \( \sum_{k=1}^{\infty} k \zeta_k(t) < 1 \) for \( t > 1 \). Therefore \( T_{gel} = 1 \).

It is important to observe that we are interested in the solution \( \zeta_k(t) \) (as in (6)) of the reduced system (5) of Smoluchowski coagulation equations as the hydrodynamic limit of \( \frac{\zeta_{k,n}(t)}{n} \) over the whole time interval \([0, \infty)\). The reason for considering \( t \in [0, \infty) \) is based
on mass conservation property in the Marcus-Lushnikov processes for all $t \geq 0$:

$$\sum_{k=1}^{\infty} k \zeta_{k,n}(t) = 1.$$  

While for the solution of (4), the mass is conserved only until $T_{gel}$, and $\sum_{k=1}^{\infty} k \zeta_k(t) < 1$ for $t > T_{gel} = 1$. As we know, in the Erdős-Rényi process, $T_{gel}$ corresponds to a time after which a single giant component emerges, and continues to absorb components of smaller size. The giant cluster dynamics is unobserved in (4), while the reduced Smoluchowski system (5) captures its influence on the dynamics of the smaller size clusters.

Indeed, in [8], P. Erdős and A. Rényi showed that the cycles are rare for a given fixed $t > 0$, and the clusters of size $k$ at time $t$ consist mainly of isolated trees of order $k$. Specifically, if $\tau_k$ denotes the number of isolated trees of order $k$, Theorem 4b in [8] asserts that

$$\lim_{n \to \infty} k E[\tau_k] = \frac{k^{k-2}t^{k-1}}{k!} e^{-kt} = \zeta_k(t)$$

and

$$\lim_{n \to \infty} \sum_{k=1}^{\infty} k E[\tau_k] = \lim_{n \to \infty} \frac{n \sum_{k=1}^{n} k E[\tau_k]}{n} = \frac{x(t)}{t}.$$  

Moreover, Theorem 9b in [8] proves the emergence of one giant component after $T_{gel} = 1$. There, if we let $\gamma_n(t)$ denote the size of the greatest component at time $t$, then

$$\lim_{n \to \infty} \frac{\gamma_n(t)}{n} = 1 - \frac{x(t)}{t}$$  

in probability.

So the dynamics of $g(t) := 1 - \sum_{k=1}^{\infty} k \zeta_k(t) = 1 - \frac{x(t)}{t}$ represents the asymptotic size of the giant component.

2.3. The length of the minimal spanning tree in $K_n$. Recall that in the construction of the Erdős-Rényi random graph model, each edge $e$ of the complete graph $K_n$ had a random variable $U_e$ associated with it. Here we consider $U_e$ to be uniform over $[0, 1]$. However, in general, various types of probability distributions are considered in the extensive literature on the topic. Now, thinking of $U_e$ as the length of the edge $e$, one can construct a minimal spanning tree on $K_n$. Let random variable $L_n$ denote the length of such minimal spanning tree. The asymptotic limit of the mean value of $L_n$ was considered in Frieze [14]. There, the results (7) and (8) from P. Erdős and A. Rényi [8] are used in proving the following limit

$$\lim_{n \to \infty} E[L_n] = \int_{0}^{\infty} \frac{x(t)}{t} dt = \sum_{k=1}^{\infty} \int_{0}^{\infty} \frac{k^{k-2}t^{k-1}}{k!} e^{-kt} dt = \zeta(3).$$
where \( \zeta(3) = \sum_{k=1}^{\infty} \frac{1}{k^3} = 1.202 \ldots \) is the value of the Riemann zeta function at 3.

Consider a coalescent process with a kernel \( K(i, j) \) for which \( T_{gel} < \infty \) has been proved. See [1, 18]. Then for a corresponding random graph model, we use the following S. Janson’s formula [17]

\[
\lim_{n \to \infty} E[L_n] = \lim_{n \to \infty} \int_0^1 E[\kappa(G(n, p))] dp - 1,
\]

where \( \kappa(G(n, p)) \) is the number of components in the Erdős-Rényi random graph \( G(n, p) \), and prove the following statement.

**Theorem 2.1.** Let \( L_n \) denote the length of the the minimal spanning tree in \( K_n \), where edge weights are independent and uniform random variables on \([0, 1]\). Then

\[
\lim_{n \to \infty} E[L_n] = \sum_{k=1}^{\infty} \int_0^{\infty} \zeta_k(t) dt,
\]

where \( \zeta_k(t) \) are the solutions (6) of the corresponding system of the reduced Smoluchowski coagulation equations (5).

Observe that the above equation (11) reproduces the result (9) of Frieze [14].

We already observed that the Marcus Lushnikov process \( \text{ML}_n(t) \) corresponding to the multiplicative coalescent process that begins with \( n \) singletons is equivalent to the cluster size dynamics in the process \( G(n, 1 - e^{-t/n}) \). Here

\[
\lim_{n \to \infty} E[L_n] = \lim_{n \to \infty} \int_0^1 E[\kappa(G(n, p))] dp - 1 = \lim_{n \to \infty} \int_0^{\infty} \frac{1}{n} E[\kappa(G(n, 1 - e^{-t/n}))] e^{-t/n} dt - 1
\]

\[
= \lim_{n \to \infty} \int_0^{\infty} \sum_{k=1}^{\infty} \frac{1}{n} E[\kappa^r(k, n, 1 - e^{-t/n})] e^{-t/n} dt - 1
\]

\[
= \lim_{n \to \infty} \int_0^{\infty} \sum_{k=1}^{\infty} \frac{E[\zeta_k(n)]}{n} e^{-t/n} dt - 1,
\]

where \( \zeta_k(t) \) are the solutions (6) of the corresponding system of the reduced Smoluchowski coagulation equations (5).
where $\kappa^{er}(k,n,p)$ is the number of components of size $k$ in $G(n,p)$ and $p = 1 - e^{-t/n}$. Therefore, one could informally calculate the limit as follows:

$$\lim_{n \to \infty} E[L_n] = \sum_{k=1}^{\infty} \int_0^\infty \zeta_k(t) dt + \lim_{n \to \infty} \int_{T_{gel}}^\infty \frac{1}{n} e^{-t/n} dt - 1$$

$$= \sum_{k=1}^{\infty} \int_0^\infty \zeta_k(t) dt + \lim_{n \to \infty} e^{-T_{gel}/n} - 1 = \sum_{k=1}^{\infty} \int_0^\infty \zeta_k(t) dt. \quad (12)$$

Here $\int_{T_{gel}}^\infty \frac{1}{n} e^{-t/n} dt$ represents the emergence of one giant component at time $T_{gel} = 1$.

Beveridge et al [5] extended S. Janson’s formula (10) to all connected graphs with i.i.d. uniform $[0,1]$ edge lengths:

$$E[L_n] = \int_0^1 E[\kappa(G(n,p))] dp - 1.$$ 

This allows us to state our general objective as follows. Consider a Marcus-Lushnikov processes equivalent to the cluster size dynamics in a general random graph model. The solutions $\zeta_k(t)$ for the corresponding reduced Smoluchowski coagulation equations are considered with $k$ in a certain index space, e.g. $k \in \mathbb{Z}_+^2 \setminus \left\{ \begin{bmatrix} 0 \\ 0 \end{bmatrix} \right\}$ in the model analyzed in Section 3. In case of a gelling kernel, the following generalization of formula (11) is proposed as a method for computing the mean length of the minimal spanning tree:

$$\lim_{n \to \infty} E[L_n] = \sum_k \int_0^\infty \zeta_k(t) d(t).$$

\[2.4. \textbf{Proof of Theorem 2.1.}\] Here we give a rigorous proof of the approach in formula (12). Note that unlike the original proof in Frieze [14], our proof will not rely on knowing the distribution of sizes and the geometry of clusters in the Erdős-Rényi process as provided in [8]. Nor will it require knowing anything about large clusters or the emergence of a unique giant component at time $T_{gel} = 1$. All that we use is the weak convergence results of Kurtz [10, 20] that we applied to the Marcus-Lushnikov processes in Section 4.

\textit{Proof.} Observe that

$$\lim_{t \to \infty} \sum_{k=1}^{\infty} k \zeta_k(t) = \lim_{t \to \infty} \sum_{k=1}^{\infty} \frac{k^{k-1}t^{k-1}}{k!} e^{-kt} = \lim_{t \to \infty} \frac{x(t)}{t} = 0. \quad (13)$$
Thus, for any given $\epsilon \in (0, 1/4)$, we can fix $T \gg T_{gel}$ so large that

$$\sum_{k=1}^\infty k\zeta_k(T) \leq \frac{\epsilon}{2}. \tag{14}$$

Notice that the above inequality (14) ties $T$ to $\epsilon$.

Fix integer $K > 0$. Since we know from Section 4 that $\lim_{n \to \infty} \frac{\zeta_k(t)}{n} = \zeta_k(t)$ a.s. over the interval $[0, T]$ for all $k \in \{1, 2, \ldots, K\}$, the probability of the complement of the event

$$Q_{K,T,n}^\epsilon := \left\{ \sum_{k=1}^K \frac{k\zeta_k(T)}{n} \leq \epsilon \right\} \tag{15}$$

is decreasing to zero as $n \to \infty$. Moreover,

$$q_{K,T}(n) := P(Q_{K,T,n}^\epsilon) = O(n^{-2})$$

by Proposition 4.3 in Subsection 4.4 as $\sum_{k=1}^K \frac{k\zeta_k(0)}{n} - \sum_{k=1}^K k\zeta_k(0) = 0$.

We will split $\int_0^\infty \sum_{k=1}^\infty E[\zeta_k(t)] e^{-t/n} dt$ as follows.

$$\int_0^\infty \sum_{k=1}^\infty E[\zeta_k(t)] e^{-t/n} dt = \int_0^T \sum_{k=1}^K E[\zeta_k(t)] e^{-t/n} dt \tag{Term I}$$

$$+ \int_0^T \sum_{k=K+1}^\infty \frac{E[\zeta_k(t)]}{n} e^{-t/n} dt \tag{Term II}$$

$$+ (1 - q_{K,T}(n)) \int_T^\infty \sum_{k=1}^K \frac{E[\zeta_k(t) \mid Q_{K,T,n}^\epsilon]}{n} e^{-t/n} dt \tag{Term III}$$

$$+ (1 - q_{K,T}(n)) \int_T^\infty \sum_{k=K+1}^\infty \frac{E[\zeta_k(t) \mid Q_{K,T,n}^\epsilon]}{n} e^{-t/n} dt \tag{Term IV}$$

$$+ q_{K,T}(n) \int_T^\infty \sum_{k=1}^\infty \frac{E[\zeta_k(t) \mid Q_{K,T,n}^\epsilon]}{n} e^{-t/n} dt \tag{Term V}$$

Next, we estimate the terms I-V in (16).
Term I. As it is proven in Section 4, \( \lim_{n \to \infty} \frac{\zeta_{k,n}(t)}{n} = \zeta_k(t) \) a.s. on \([0, T]\) for all \( k = 1, 2, \ldots, K \). Therefore,

\[
\lim_{n \to \infty} \int_0^T \sum_{k=1}^{K} \frac{E[\zeta_{k,n}(t)\mid Q_{K,T,n}]}{n} e^{-t/n} dt = \sum_{k=1}^{K} \int_0^T \zeta_k(t) dt.
\]

Term II. Observe that,

\[
\sum_{k=K+1}^{\infty} \frac{\zeta_{k,n}(t)}{n} \leq \frac{1}{Kn} \sum_{k=K+1}^{\infty} k\zeta_{k,n}(t) = \frac{1}{K} \left( 1 - \sum_{k=1}^{K} \frac{k\zeta_{k,n}(t)}{n} \right) \leq \frac{1}{K}.
\]

Thus,

\[
\int_0^T \sum_{k=K+1}^{\infty} \frac{E[\zeta_{k,n}(t)\mid Q_{K,T,n}]}{n} e^{-t/n} dt = O\left( \frac{T}{K} \right)
\]

regardless of the value of \( n > 0 \).

Term III. Recall that in the theory of Marcus-Lushnikov processes the gel is the set of all “large” clusters. By analogy, we define the \( K \)-gel to be the collection of all clusters of mass bigger than \( K \). Let \( M_{K\text{-gel}}(t) \) denote the total mass of all clusters in the \( K \)-gel at time \( t \geq 0 \).

Now, conditioning on the event \( Q_{K,T,n} \), the mass of the \( K \)-gel is \( M_{K\text{-gel}}(t) \geq (1 - \epsilon)n \) for all \( t \geq T \). Thus each cluster not in \( K \)-gel will be gravitating toward the \( K \)-gel with the rate of at least \( \frac{M_{K\text{-gel}}(t)}{n} \geq 1 - \epsilon \). Consider a cluster that was not in \( K \)-gel at time \( T \). Let \( T + L \) be the time it becomes a part of the \( K \)-gel. Then, its contribution to the integral \( \int_0^T \sum_{k=1}^{K} \frac{E[\zeta_{k,n}(t)\mid Q_{K,T,n}]}{n} e^{-t/n} dt \) is at most

\[
\int_T^\infty \frac{E[1_{[T,T+L]}(t)\mid Q_{K,T,n}]}{n} e^{-t/n} dt \leq \frac{\int_T^\infty E[1_{[T,T+L]}(t)\mid Q_{K,T,n}] dt}{n} e^{-T/n} = \frac{E[L\mid Q_{K,T,n}]}{n} e^{-T/n} \leq \frac{1}{(1 - \epsilon)n},
\]

where

\[
1_A = \begin{cases} 
1 & \text{if } t \in A \\
0 & \text{if } t \notin A
\end{cases}
\]

The number of clusters not in \( K \)-gel at time \( t \geq T \) is

\[
\sum_{k=1}^{K} \zeta_{k,n}(t) \leq \sum_{k=1}^{K} k\zeta_{k,n}(t) \leq \epsilon n.
\]
Therefore,
\[
\int_0^\infty \sum_{k = 1}^K \frac{E[\zeta_{k,n}(t) | Q_{K,T,n}^\epsilon]}{n} e^{-t/n} dt \leq \frac{\epsilon n}{(1 - \epsilon)n} e = \frac{\epsilon}{1 - \epsilon} < 2\epsilon.
\]

**Term IV.** We let \( \mathcal{C} = \{C_1, C_2, C_3, \ldots, C_M\} \) denote the set of all clusters that ever exceeded mass \( K \) in the whole history of the process \( \{M_{L_n(t)}\}_{t \in [0, \infty)} \). There are less than \( n/K \) such clusters, i.e. \( M < n/K \). For each \( C_i \), the emergence time \( a_i \) is the time when a pair of clusters of mass not exceeding \( K \) mergers into a new cluster \( C_i \) of mass greater than \( K \). We enumerate these clusters in the order they emerge.

Let \( M_i(t) \) denote the mass of cluster \( C_i \) at time \( t \). Consider a pair of clusters, \( C_i \) and \( C_j \), coexisting in the \( K-gel \) at time \( t \), each of mass smaller than \( n/2 \). We split their merger rate into two by saying that \( C_i \) absorbs \( C_j \) with rate \( \frac{1}{2n} M_i(t) M_j(t) \), and \( C_j \) absorbs \( C_i \) with rate \( \frac{1}{2n} M_i(t) M_j(t) \). In other words, \( C_i \) and \( C_j \) merge with rate \( \frac{1}{n} M_i(t) M_j(t) \), and which one of the two clusters absorbs the other is decided with a toss of an independent fair coin.

There is a finite stopping time
\[
t^* = \min\{t \geq 0 : \exists C_i \in \mathcal{C} \text{ with } M_i(t) \geq n/2\}
\]
when a cluster \( C_{i^*} \) has its mass \( M_{i^*}(t^*) \geq n/2 \). After \( t^* \), the rules of interactions of cluster \( C_{i^*} \) with the other clusters in \( \mathcal{C} \) change as follows. For \( t > t^* \), \( C_{i^*} \) absorbs \( C_j \) with rate \( \frac{1}{n} M_{i^*}(t) M_j(t) \), while \( C_{i^*} \) itself cannot be absorbed by any other cluster in \( \mathcal{C} \).

Let \( b_i \) denote the time when cluster \( C_i \) is absorbed by another cluster in collection \( \mathcal{C} \). Naturally, there will be only one survivor \( C_{i^*} \) with \( b_{i^*} = \infty \). Let \( J_i = [a_i, b_i] \cap [T, \infty) \) denote the lifespan of cluster \( C_i \). Note that a cluster \( C_i \) from the set \( \mathcal{C} \) existing at time \( t \in [a_i, b_i] \) is absorbed into one of the clusters in the \( K-gel \) with the total instantaneous rate of
\[
\lambda_i(t) \geq \frac{1}{2n} M_i(t) \left( M_{K-gel}(t) - M_i(t) \right).
\]
Conditioning on the event \( Q_{K,T,n}^\epsilon \) defined in (15), we have that if \( M_i(t) < n/2 \) for \( t \in J_i \), then the rate of absorption of \( C_i \) into the \( K-gel \) is
\[
\lambda_i(t) \geq \frac{1}{2n} M_i(t) \left( 1 - \epsilon \right) n - \frac{1}{2} n \geq \frac{1}{2n} M_i(t) \left( \frac{3}{4} n - \frac{1}{2} n \right) \geq \frac{1}{8} M_i(t) > \frac{K}{8}.
\]

Next,
\[
\int_0^\infty \sum_{k = K+1}^\infty \frac{E[\zeta_{k,n}(t) | Q_{K,T,n}^\epsilon]}{n} e^{-t/n} dt = \int_0^\infty \frac{1}{n} e^{-t/n} dt + \mathcal{E}
\]
where \( \int_{T}^{\infty} e^{-t/n} dt \) is due to the event \( Q_{K,T,n}^{i} \) which guarantees the existence of at least one component from \( C \) in the \( K\text{-gel} \) for all \( t \in [T, \infty) \) and the second term \( \mathcal{E} \) is responsible for all the times \( t \geq T \) when the number of clusters in the \( K\text{-gel} \) is greater than one. The term \( \mathcal{E} \) is bounded as follows

\[
\mathcal{E} \leq \int_{T}^{\infty} \frac{E \left[ \left. \sum_{i: i \neq i^{*}} 1_{J_{i}(t)} \right| Q_{K,T,n}^{i} \right] e^{-t/n} dt}{n}.
\]

Now, each cluster \( C_{i} \) is gravitating towards the rest of the \( K\text{-gel} \) with the rate of at least \( K/8 \). Thus, for each \( i \neq i^{*} \),

\[
\int_{T}^{\infty} \frac{E \left[ \left. 1_{J_{i}(t)} \right| Q_{K,T,n}^{i} \right] e^{-t/n} dt}{n} \leq \frac{E \left[ |J_{i}| \right| Q_{K,T,n}^{i} \right]}{n} e^{-T/n} \leq \frac{8}{nK}.
\]

Hence, since the cardinality of set \( C \) is \( M < n/K \),

\[
\mathcal{E} < \frac{n}{K} \cdot \frac{8}{nK} = \frac{8}{K^{2}},
\]

and from (17), we obtain

\[
\int_{T}^{\infty} \sum_{k=K+1}^{\infty} \frac{E[\zeta_{k,n}(t) \mid Q_{K,T,n}^{k}]}{n} e^{-t/n} dt = 1 + O(K^{-2}) + O\left(\frac{T}{n}\right) \quad \text{as } n \to \infty,
\]

where the term \( O(K^{-2}) \) does not depend on the value of \( n > 0 \).

**Term V.** Here

\[
q_{K,T}^{i}(n) \int_{T}^{\infty} \sum_{k=1}^{\infty} \frac{E[\zeta_{k,n}(t) \mid Q_{K,T,n}^{k}]}{n} e^{-t/n} dt \leq nq_{K,T}^{i}(n) \int_{T}^{\infty} \frac{1}{n} e^{-t/n} dt \leq nq_{K,T}^{i}(n) = O(n^{-1}).
\]

Finally, by putting together the analysis in **Terms I-V** in the equation (16), we obtain for a given fixed \( \epsilon \in (0, 1/4) \), sufficiently large fixed \( T \gg T_{gel} \) satisfying (14), and arbitrarily large \( K \),

(18)

\[
\int_{0}^{\infty} \sum_{k=1}^{\infty} \frac{E[\zeta_{k,n}(t)]}{n} e^{-t/n} dt = \sum_{k=1}^{K} \int_{0}^{T} \zeta_{k}(t) dt + 1 + O\left(\frac{T}{K}\right) + O(K^{-2}) + O(\epsilon) + O\left(\frac{T}{n}\right) + O(n^{-1}),
\]
which, when we increase $n$ to infinity will yield
\[
\limsup_{n \to \infty} \left| \int_0^\infty \sum_{k=1}^\infty \frac{E[\zeta_{k,n}(t)]}{n} e^{-t/n} \, dt - \sum_{k=1}^\infty \int_0^\infty \zeta_k(t) \, dt - 1 \right| = \sum_{k=K+1}^\infty \int_0^T \zeta_k(t) \, dt + \sum_{k=1}^\infty \int_T^\infty \zeta_k(t) \, dt + O\left(\frac{T}{K}\right) + O(K^{-2}) + O(\epsilon).
\]
Consequently, taking $\limsup_{K \to \infty}$, we obtain
\[
\limsup_{n \to \infty} \left| \int_0^\infty \sum_{k=1}^\infty \frac{E[\zeta_{k,n}(t)]}{n} e^{-t/n} \, dt - \sum_{k=1}^\infty \int_0^\infty \zeta_k(t) \, dt - 1 \right| = \sum_{k=1}^\infty \int_0^\infty \zeta_k(t) \, dt + O(\epsilon).
\]
Finally, formula (14) guarantees that decreasing $\epsilon$ down to zero will propel $T$ to $+\infty$, and
\[
\lim_{n \to \infty} \int_0^\infty \sum_{k=1}^\infty \frac{E[\zeta_{k,n}(t)]}{n} e^{-t/n} \, dt = \sum_{k=1}^\infty \int_0^\infty \zeta_k(t) \, dt + 1.
\]
Thus we confirmed formula (11) for the case of the multiplicative coalescent process. □

2.5. Gelation. Here, we would like to summarize the main results regarding the gelation phenomenon. Consider a general system of Smoluchowski coagulation equations with a positive symmetric kernel $K(i,j) = K(j,i) > 0$,
\[
\frac{d}{dt} \zeta_j = -\zeta_j \sum_{j=1}^\infty K(i,j) \zeta_i + \frac{1}{2} \sum_{i=1}^{j-1} K(i,j-i) \zeta_i \zeta_{j-i} \quad (k = 1, 2, \ldots) \text{ with } \zeta_j(0) = \delta_{1,j}.
\]
Then, following [23], we use the above Smoluchowski coagulation equations to obtain
\[
\frac{d}{dt} \sum_{j=1}^\infty j \zeta_j = \sum_{j=1}^\infty j \frac{d}{dt} \zeta_j = -\sum_{i,j=1}^\infty jK(i,j) \zeta_j \zeta_i + \frac{1}{2} \sum_{j=1}^\infty \sum_{i=1}^{j-1} (i + (j-i)) K(i,j-i) \zeta_i \zeta_{j-i}
\]
\[
= -\sum_{i,j=1}^\infty jK(i,j) \zeta_j \zeta_i + \frac{1}{2} \sum_{i,j=1}^\infty (i + j) K(i,j) \zeta_i \zeta_j = 0
\]
provided convergence of $\sum_{i,j=1}^\infty jK(i,j) \zeta_j \zeta_i$. Therefore, letting
\[
T_{gel} := \sup \left\{ t > 0 : \sum_{i,j=1}^\infty jK(i,j) \zeta_j(s) \zeta_i(s) \text{ converges for all } s \in [0,t] \right\},
\]
we have $\frac{d}{dt} \sum_{j=1}^\infty j \zeta_j = 0$ for $t \in [0, T_{gel})$, which in turn implies $\sum_{j=1}^\infty j \zeta_j(t) = 1$ for $t \in [0, T_{gel})$. 

The question whether $T_{gel} < \infty$ is the question of whether gelation phenomenon occurs in a given system of Smoluchowski equations. The first mathematical proof of gelation was produced in McLeod [23] for the multiplicative kernel. Historically, that happened around the time when the formation of a giant cluster was proved by P. Erdős and A. Rényi [8]. The overlap in mathematical formulas obtained in the two papers, [23] and [8], representing the two different branches of mathematics is quite remarkable. The work of finding a mathematically solid proof of gelation phenomenon for other conjectured gelling kernels began fifteen years later with a work by Lushnikov [21]. It continued with publications of Ziff [32], Ernst et al. [9], van Dongen and Ernst [30], and many other mathematicians and mathematical physicists. In Spouge [29], the gelation is demonstrated numerically for the general bilinear kernel $K_{ij} = A + B(i + j) + C_{ij}$. While in Aldous [1], the gelation is proved for $K_{ij} = 2(ij)\gamma (i + j)\gamma - i\gamma - j\gamma$, where $\gamma \in (1, 2)$. There $\gamma = 2$ would correspond to the multiplicative kernel for which, as we know, gelatin also occurs. Jeon [18] proved that complete and instantaneous gelation occurs if $K_{ij} \geq ij\psi(i, j)$, where $\psi(i, j)$ is a function increasing in both variables, $i$ and $j$, such that $\sum_{j=1}^{\infty} \frac{1}{\psi(i, j)} < \infty$ for all $i$. This includes $K_{ij} = (ij)^{\alpha}$, $\alpha > 1$, as a primary example.

3. Erdős-Rényi process on $K_{\alpha n, \beta n}$ and cross-multiplicative coalescent

In this section, we study an Erdős-Rényi process on the complete bipartite graph with partitions of size $\alpha n + o(\sqrt{n})$ and $\beta n + o(\sqrt{n})$ and the corresponding novel cross-multiplicative coalescent process with kernel $K((i_1, j_1), (i_2, j_2)) = i_1j_2 + i_2j_1$. We will use the weak convergence results from Section 4 to prove the main results of this paper, Theorem 3.2 and Theorem 3.3.

For $\alpha, \beta > 0$, consider two integer valued functions, $\alpha[n] = \alpha n + o(\sqrt{n})$ and $\beta[n] = \beta n + o(\sqrt{n})$. Consider an Erdős-Rényi random graph process on the bipartite graph $K_{\alpha[n], \beta[n]}$ with $\alpha[n]$ vertices on the left side and $\beta[n]$ vertices on the right side. There each edge $e$ of $\sim \alpha\beta n^2$ edges has an associated random variable $U_e$. The random variables $U_e$ are assumed to be independent and uniform over $[0, 1]$. For the time parameter $p \in [0, 1]$, an edge $e$ is considered open if $U_e \leq p$. Erdős-Rényi random graph $G(n, p)$ will consist of all $n$ vertices and all open edges at time $p$.

Next, consider a coalescent process corresponding to an Erdős-Rényi random graph process on $K_{\alpha[n], \beta[n]}$. Specifically, let each cluster connecting $i_1$ vertices on the left side of the bipartite graph with $i_2$ vertices on the right side of the bipartite graph be assigned a two-dimensional weight vector $i = [i_1 \ i_2]$. There $i_1, i_2 \geq 0$ and $i_1 + i_2 > 0$. Define the coalescence kernel as follows. For any pair of clusters with weight vectors $i = [i_1 \ i_2]$ and
\( j = \begin{bmatrix} j_1 \\ j_2 \end{bmatrix} \), let
\begin{equation}
K(i, j) := i_1 j_2 + i_2 j_1.
\end{equation}

The coalescent process begins with \( \alpha[n] + \beta[n] \) singletons, of which \( \alpha[n] \) singletons are of weight \( \begin{bmatrix} 1 \\ 0 \end{bmatrix} \) and the other \( \beta[n] \) singletons are of weight \( \begin{bmatrix} 0 \\ 1 \end{bmatrix} \). There, each pair of clusters of respective weights \( i \) and \( j \) would coalesce into a cluster of weight \( i + j \) with rate \( K(i, j)/n \). The last merger will create a cluster of weight \( \alpha[n] \beta[n] \). We will call this a \textbf{cross-multiplicative coalescent process}. Note that this cross-multiplicative coalescent process represents the cluster dynamics of the above Erdős-Rényi random graph process on the bipartite graph \( K_{\alpha[n], \beta[n]} \) under the time change \( p = 1 - e^{-t/n} \).

3.1. \textbf{Smoluchowski coagulation equations}. Consider the Marcus-Lushnikov process \( \text{ML}_n(t) \) that keeps track of cluster counts in the above defined cross-multiplicative coalescent process that begins with the \( \alpha[n] + \beta[n] \) singletons of the two types, \( \alpha[n] \) of weight \( \begin{bmatrix} 1 \\ 0 \end{bmatrix} \) and \( \beta[n] \) of weight \( \begin{bmatrix} 0 \\ 1 \end{bmatrix} \). Specifically, let \( \zeta_{i_1, i_2}^{[n]}(t) \) denote the number of the components in the cross-multiplicative coalescent process of weight \( i = \begin{bmatrix} i_1 \\ i_2 \end{bmatrix} \) at time \( t \). Then \( \text{ML}_n(t) \) is the infinite-dimensional process with coordinates \( \zeta_{i_1, i_2}^{[n]}(t) \), i.e.
\begin{equation}
\text{ML}_n(t) = \left( \zeta_{i_1, i_2}^{[n]}(t) \right)_{i_1, i_2}
\end{equation}
with the starting values \( \zeta_{i_1, i_2}^{[n]}(0) = \alpha \delta_{i_1, 0} \delta_{i_2, 1} + \beta \delta_{0, i_1} \delta_{1, i_2} \).

The Smoluchowski coagulation equations for the Marcus-Lushnikov process \( \text{ML}_n(t) \) with bipartite multiplicative kernel are written as follows:
\begin{equation}
\frac{d}{dt} \zeta_{i_1, i_2}(t) = -\sum_{j_1, j_2} \zeta_{i_1, i_2}(t) (i_1 j_2 + i_2 j_1) \zeta_{j_1, j_2}(t) + \frac{1}{2} \sum_{\ell_1, k_1: \ell_1 + k_1 = i_1, \ell_2, k_2: \ell_2 + k_2 = i_2} (\ell_1 k_2 + \ell_2 k_1) \zeta_{\ell_1, \ell_2}(t) \zeta_{k_1, k_2}(t)
\end{equation}
with the initial conditions \( \zeta_{i_1, i_2}(0) = \alpha \delta_{i_1, 0} \delta_{i_2, 1} + \beta \delta_{0, i_1} \delta_{1, i_2} \).

A reduced system of differential equation corresponding to the above Smoluchowski coagulation equations (20) will be given in (21). It will take into account the mass conservation property of the above Marcus-Lushnikov process \( \text{ML}_n(t) \), and therefore will represent the smaller cluster dynamics over the whole time interval \([0, \infty)\).
First, we notice that here the initial total mass is \( \sum_{i_1,i_2} (i_1 + i_2) \zeta_{i_1,j_1}(0) = \alpha + \beta \). Moreover, the initial total ‘left mass’ is \( \sum_{i_1,i_2} i_1 \zeta_{i_1,i_2}(0) = \alpha \) and the initial total ‘right mass’ is \( \sum_{i_1,i_2} i_2 \zeta_{i_1,i_2}(0) = \beta \).

Next, we consider the rate of change for the total left mass and the total right mass, and use (20) to obtain

\[
\frac{d}{dt} \sum_{i_1,i_2} i_1 \zeta_{i_1,i_2}(t) = - \sum_{i_1,i_2,j_1,j_2} i_1(i_1j_2 + i_2j_1) \zeta_{i_1,i_2}(t) \zeta_{j_1,j_2}(t) + \frac{1}{2} \sum_{\ell_1,k_1,\ell_2,k_2} (\ell_1 + k_1)(\ell_1k_2 + \ell_2k_1) \zeta_{\ell_1,\ell_2}(t) \zeta_{k_1,k_2}(t) = 0
\]

and

\[
\frac{d}{dt} \sum_{i_1,i_2} i_2 \zeta_{i_1,i_2}(t) = - \sum_{i_1,i_2,j_1,j_2} i_2(i_1j_2 + i_2j_1) \zeta_{i_1,i_2}(t) \zeta_{j_1,j_2}(t) + \frac{1}{2} \sum_{\ell_1,k_1,\ell_2,k_2} (\ell_2 + k_2)(\ell_1k_2 + \ell_2k_1) \zeta_{\ell_1,\ell_2}(t) \zeta_{k_1,k_2}(t) = 0
\]

whenever \( \sum_{i_1,i_2} (i_1 + i_2)^2 \zeta_{i_1,i_2}(t) \) converges. Thus in order to establish whether the kernel defined in (19) is a gelling kernel, we need to consider whether

\[
T_{gel} := \sup \{ t > 0 : \sum_{i_1,i_2} (i_1 + i_2)^2 \zeta_{i_1,i_2}(t) \text{ converges} \}
\]

is finite.

Here, for \( t < T_{gel} \), \( \sum_{j_1,j_2} j_1 \zeta_{j_1,j_2}(t) = \alpha \) and \( \sum_{j_1,j_2} j_2 \zeta_{j_1,j_2}(t) = \beta \). Therefore, for any \( i_1 \) and \( i_2 \), \( \sum_{j_1,j_2} (i_1j_2 + i_2j_1) \zeta_{j_1,j_2}(t) = \beta i_1 + \alpha i_2 \). Thus we can consider the following reduced Smoluchowski coagulation equations:

\[
(21) \quad \frac{d}{dt} \zeta_{i_1,i_2}(t) = -(\beta i_1 + \alpha i_2) \zeta_{i_1,i_2}(t) + \frac{1}{2} \sum_{\ell_1,k_1 : \ell_1+k_1=i_1, \ell_2,k_2 : \ell_2+k_2=i_2} (\ell_1k_2 + \ell_2k_1) \zeta_{\ell_1,\ell_2}(t) \zeta_{k_1,k_2}(t)
\]

with the initial conditions \( \zeta_{i_1,i_2}(0) = \alpha \delta_{i_1,i_2} + \beta \delta_{0,i_1} \delta_{0,i_2} \). Once again, the solutions of Smoluchowski coagulation system (20) and the above reduced Smoluchowski coagulation system (21) will match up until \( T_{gel} \). Consecutively, the solution (25) of the reduced Smoluchowski system of equations (21) will be used in Subsection 3.3 in establishing the finiteness of the gelation time and for finding its value, \( T_{gel} \).
The following hydrodynamic limit is proven in the equation (44) of Subsection 4.3. Fix a pair of positive integers $K_1$ and $K_2$, and a real $T > 0$. Then,

$$
\lim_{n \to \infty} \sup_{s \in [0,T]} n^{-1} \left| n\zeta_{i_1,i_2}(s) - \zeta_{i_1,i_2}(s) \right| = 0 \quad \text{a.s.}
$$

for all $i_1, i_2 \geq 1$, where $\zeta_{i_1,i_2}(t)$ solves the reduced Smoluchowski coagulation system (21). Consequently,

$$
\lim_{n \to \infty} \sup_{s \in [0,T]} n^{-1} \sum_{1 \leq i_1 \leq K_1} \sum_{1 \leq i_2 \leq K_2} \zeta_{i_1,i_2}(s) - \sum_{1 \leq i_1 \leq K_1} \sum_{1 \leq i_2 \leq K_2} \zeta_{i_1,i_2}(s) = 0 \quad \text{a.s.}
$$

3.2. Solution $\zeta_{i_1,i_2}(t)$ of (21). Observe that as it was the case with the multiplicative coalescent, we will consider the above reduced system of differential equations (21) over the whole time interval $[0, \infty)$ because the mass conservation property

$$
\sum_{i_1,i_2} i_1 \zeta_{i_1,i_2}(t) = \alpha \quad \text{and} \quad \sum_{i_1,i_2} i_2 \zeta_{i_1,i_2}(t) = \beta
$$

in the Marcus-Lushnikov process $ML_n(t)$ holds for all $t \in [0, \infty)$. Thus, while the solutions to (20) and (21) are identical over $[0, T_{gel})$, the reduced system (21) continues to reflect the dynamics of the smaller clusters even after $T_{gel}$.

Next, we want to find the solution $\zeta_{i_1,i_2}(t)$ of reduced system (21) for all $t \geq 0$. Here we observe that $\zeta_{1,0}(t) = \alpha e^{-\beta t}$ and $\zeta_{0,1}(t) = \beta e^{-\alpha t}$, and extend the approach of McLeod [23] by considering the solutions of the following form

$$
\zeta_{i_1,i_2}(t) = \alpha^{i_1} \beta^{i_2} S_{i_1,i_2} e^{-(\beta i_1 + \alpha i_2) t} i_1^{i_1} i_2^{i_2} - 1
$$

and plugging them into the equation (21). After cancelations, we arrive with the following recursion

$$
(i_1 + i_2 - 1) S_{i_1,i_2} = \frac{1}{2} \sum_{\ell_1,k_1: \ell_1 + k_1 = i_1, \ell_2,k_2: \ell_2 + k_2 = i_2} (\ell_1 k_2 + \ell_2 k_1) S_{\ell_1,\ell_2} S_{k_1,k_2}
$$

with the initial conditions $S_{i,0} = S_{0,i} = \delta_{i,i}$. Here, $S_{i_1,i_2} = S_{i_2,i_1}$.

Here, we managed to solve the problem of finding $S_{i_1,i_2}$ combinatorially.

**Proposition 3.1.** The system of equations (23) with the initial conditions $S_{i,0} = S_{0,i} = \delta_{i,i}$ has the following unique solution

$$
S_{i_1,i_2} = \frac{i_2^{i_2 - 1} i_1^{i_1 - 1}}{i_1! i_2!}.
$$
Proof. In Theorem 1.1(3) of [16], F. Huang and B. Liu generalize the Abel’s binomial theorem as follows:

\[
\sum_{k_1=0}^{i_1} \sum_{k_2=0}^{i_2} \binom{i_1}{k_1} \binom{i_2}{k_2} (v + z i_1 - z k_1)^{i_2 - k_2} (u + z k_2)^{i_1 - k_1 - 1} = \frac{[u v - i_1 i_2 z^2] u^{i_1 - 1} v^{i_2 - 1}}{(v + i_1 z)(u + i_2 z)}
\]

Then, we use (24) with \( z = -1 \) to confirm our candidate solution satisfies (23) by plugging it into the right hand side of (23) as follows.

\[
\frac{1}{2} \sum_{\ell_1, k_1: \ell_1 + k_1 = i_1, \ell_2, k_2: \ell_2 + k_2 = i_2} (\ell_1 k_2 + 2 k_1) S_{\ell_1, \ell_2} S_{k_1, k_2} = \sum_{\ell_1, k_1: \ell_1 + k_1 = i_1, \ell_2, k_2: \ell_2 + k_2 = i_2} \ell_1 k_2 \ell_2 \ell_1^i k_1^2 k_2^i \ell_1 \ell_2 \ell_1^i k_1^2 k_2^i
\]

\[
= \frac{1}{i_1! i_2!} \sum_{k_1: 0 \leq k_1 \leq i_1, k_2: 0 \leq k_2 \leq i_2, (k_1, k_2) \neq (0,0), (i_1, i_2)} \binom{i_1}{k_1} \binom{i_2}{k_2} (v - i_1 + k_1)^{k_2 - 1} (i_1 - k_1)^{i_2 - k_2} k_1^i k_2^i (u - k_2)^{i_1 - k_1 - 1}
\]

\[
= \frac{1}{i_1! i_2!} \lim_{u \to i_1, v \to i_2} \left\{ \sum_{k_1=0}^{i_1} \sum_{k_2=0}^{i_2} \binom{i_1}{k_1} \binom{i_2}{k_2} (v - i_1 + k_1)^{k_2 - 1} (i_1 - k_1)^{i_2 - k_2} k_1^i k_2^i (u - k_2)^{i_1 - k_1 - 1}
\]

\[
= \frac{1}{i_1! i_2!} \lim_{u \to i_1, v \to i_2} \left\{ \frac{[u v - i_1 i_2 z^2] u^{i_1 - 1} v^{i_2 - 1}}{(v - i_1)(u - i_2)} - \frac{i_1^{i_2} u^{i_1 - 1}}{v - i_1} - \frac{i_2^{i_2} v^{i_2 - 1}}{u - i_2} \right\}
\]

\[
= \frac{1}{i_1! i_2!} \lim_{u \to i_1, v \to i_2} \left\{ \frac{i_1^{i_2 - 1} u^{i_1 - 1}}{v - i_1} + \frac{i_2^{i_2 - 1} v^{i_2 - 1}}{u - i_2} - \frac{i_2^{i_2 - 1} u^{i_1 - 1}}{v - i_1} - \frac{i_1^{i_2} v^{i_2 - 1}}{u - i_2} \right\}
\]
Hence,
\[
\frac{1}{2} \sum_{\ell_1, k_1: \ell_1 + k_1 = i_1, \ell_2, k_2: \ell_2 + k_2 = i_2} (\ell_1 k_2 + \ell_2 k_1) S_{\ell_1, \ell_2} S_{k_1, k_2} = \frac{1}{i_1! i_2!} \lim_{u \to i_1} \left\{ i_1 u^{i_1 - 1} v^{i_2 - 1} - t_1^{i_1 - 1} + v^{i_2 - 1} u^{i_1 - i_2} \right\}
\]
\[
= \frac{1}{i_1! i_2!} \left( (i_2 - 1) \cdot (i_1 i_2 - 1) + i_1 \cdot (i_2 - 1) \right)
\]
\[
= (i_1 + i_2 - 1) S_{i_1, i_2}
\]
thus completing the proof. \(\square\)

The solution of equations (21) follows from (22) and Proposition 3.1,
\[
(25) \quad \zeta_{i_1, i_2}(t) = \frac{\alpha_i \beta_i}{i_1! i_2!} e^{-(\beta_i + \alpha_i) t} t_{i_1, i_2}.
\]

3.3. Gelation in \(K_{\alpha, \beta n}\). Next, we prove the finiteness of the gelation time
\[
T_{gel} := \sup \left\{ t > 0 : \sum_{i_1, i_2} (i_1 + i_2)^2 \zeta_{i_1, i_2}(t) \right\}
\]
From equation (25) we know that \(T_{gel}\) is the radius of convergence for the series
\[
\sum_{i_1, i_2} (i_1 + i_2)^2 \zeta_{i_1, i_2}(t) = \sum_{i_1, i_2} (i_1 + i_2)^2 \frac{\alpha_i \beta_i}{i_1! i_2!} e^{-(\beta_i + \alpha_i) t} t_{i_1, i_2}.
\]
Here, by Stirling's approximation, \(T_{gel}\) solves \(1 - (\alpha \land \beta) t + \ln((\alpha \lor \beta) t) = 0\).
We also notice that the mass of the system in (21) is conserved until \(T_{gel}\), after which it begins to dissipate, i.e.
\[
\begin{cases}
\sum_{i_1, i_2} (i_1 + i_2) \zeta_{i_1, i_2}(t) = \alpha + \beta & \text{if } t \leq T_{gel} \\
\sum_{i_1, i_2} (i_1 + i_2) \zeta_{i_1, i_2}(t) < \alpha + \beta & \text{if } t > T_{gel}.
\end{cases}
\]

3.4. The length of the minimal spanning tree on \(K_{\alpha, \beta n}\) via \(\zeta_{i_1, i_2}(t)\). Once again, let us consider \(U_e\) to be the length of the edge \(e\). Then one can construct a minimal spanning tree on \(K_{\alpha, \beta n}\). Let random variable \(L_n\) denote the length of such minimal spanning tree. We want to represent the asymptotic limit of the mean value of \(L_n\) via \(\zeta_{i_1, i_2}(t)\).

For a random graph process \(G(n, p)\) over \(K_{\alpha, \beta n}\), Lemma 1 in Beveridge et al [5] implies
\[
(26) \quad E[L_n] = \int_0^1 E[\kappa(G(n, p))] dp - 1,
\]
where \( \kappa(G(n, p)) \) is the number of components in the random graph process \( G(n, p) \) at time \( p \). This will be used in establishing the following theorem that will be proved in Subsection 3.5.

**Theorem 3.2.** Let \( \alpha, \beta > 0 \) and \( L_n = L_n(\alpha, \beta) \) be the length of a minimal spanning tree on a complete bipartite graph \( K_{\alpha[n], \beta[n]} \) with partitions of size

\[
\alpha[n] = \alpha n + o(\sqrt{n}) \quad \text{and} \quad \beta[n] = \beta n + o(\sqrt{n})
\]

and independent uniform edge weights over \([0, 1]\). Then

\[
\lim_{n \to \infty} E[L_n] = \sum_{i_1, i_2} \int_0^{\infty} \zeta_{i_1, i_2}(t) d(t).
\]

where \( \zeta_{i_1, i_2}(t) \) indexed by \( \mathbb{Z}_+^2 \setminus \{(0,0)\} \) is the solution of the reduced Smoluchowski coagulation system (21) with the initial conditions \( \zeta_{i_1, i_2}(0) = \alpha \delta_{i_1, i_2} + \beta \delta_{i_1, i_2} \).

Observe that if we plug-in the solutions (22) of the reduced system of Smoluchowski coagulation equations (21) into the right hand side of (27), we get

\[
\sum_{i_1, i_2} \int_0^{\infty} \zeta_{i_1, i_2}(t) d(t) = \frac{\alpha}{\beta} + \frac{\beta}{\alpha} + \sum_{i_1 \geq 1; \ i_2 \geq 1} \alpha i_1 \beta i_2 S_{i_1, i_2} \int_0^{\infty} t^{i_1 + i_2 - 1} e^{-(\beta i_1 + \alpha i_2)t} dt
\]

\[
= \frac{\alpha}{\beta} + \frac{\beta}{\alpha} + \sum_{i_1 \geq 1; \ i_2 \geq 1} \frac{\alpha i_1 \beta i_2 S_{i_1, i_2}}{(\beta i_1 + \alpha i_2)^{i_1 + i_2}} (i_1 + i_2 - 1)!
\]

\[
= \gamma + \frac{1}{\gamma} \sum_{i_1 \geq 1; \ i_2 \geq 1} \frac{\gamma i_1 S_{i_1, i_2}}{(i_1 + \gamma i_2)^{i_1 + i_2}} (i_1 + i_2 - 1)!
\]

with \( \gamma = \frac{\alpha}{\beta} \).

Next, by combining Proposition 3.1 with (28) we obtained the following main theorem.

**Theorem 3.3.** Let \( \alpha, \beta > 0 \), \( \gamma = \alpha/\beta \), and \( L_n = L_n(\alpha, \beta) \) be the length of a minimal spanning tree on a complete bipartite graph \( K_{\alpha[n], \beta[n]} \) with partitions of size

\[
\alpha[n] = \alpha n + o(\sqrt{n}) \quad \text{and} \quad \beta[n] = \beta n + o(\sqrt{n})
\]

and independent uniform edge weights over \([0, 1]\). Then the limiting mean length of the minimal spanning tree is

\[
\lim_{n \to \infty} E[L_n] = \gamma + \frac{1}{\gamma} \sum_{i_1 \geq 1; \ i_2 \geq 1} \frac{(i_1 + i_2 - 1)! \gamma i_1 i_2^{i_1 - 1} i_2^{i_2 - 1}}{i_1! i_2! (i_1 + \gamma i_2)^{i_1 + i_2}}.
\]

Theorem 3.3 is consistent with [15], where it was shown that for \( \alpha = \beta \), \( \lim_{n \to \infty} E[L_n] = 2\zeta(3) \). Indeed, we have the following Corollary reproducing the results in [15]. Observe however that for \( \alpha \neq \beta \) the bipartite graph is irregular and the results in Frieze and McDiarmid [15] no longer apply.
Corollary 3.4. If $\gamma = 1$, then

$$\lim_{n \to \infty} E[L_n] = 2\zeta(3).$$

Proof. Abel’s binomial theorem [6, 27] states that

$$\sum_{k=0}^{n} \binom{n}{k} (x-kz)^{k-1}(y+kz)^{n-k} = x^{-1}(x+y)^n.$$ 

Plugging-in $x = nz \neq 0$, $y = 0$, and $i = n - k$, we obtain

$$\sum_{i=0}^{n} \binom{n}{i} i^{n-i-1}(n-i)^i = n^{n-1}$$

and therefore,

$$\sum_{i_1,i_2: \; i_1 + i_2 = n} i_1 S_{i_1,i_2} = \sum_{i_1,i_2: \; i_1 + i_2 = n} \frac{i_1^{i_1-1} i_2^{i_2-1}}{i_1! i_2!} = \frac{n^{n-1}}{n!}.$$ 

Hence,

$$n \cdot \sum_{i_1,i_2: \; i_1 + i_2 = n} S_{i_1,i_2} = \sum_{i_1,i_2: \; i_1 + i_2 = n} (i_1 + i_2) S_{i_1,i_2} = 2 \sum_{i_1,i_2: \; i_1 + i_2 = n} i_1 S_{i_1,i_2} = 2 \frac{n^{n-1}}{n!}$$

and

$$\sum_{i_1,i_2: \; i_1 + i_2 = n} S_{i_1,i_2} = 2 \frac{n^{n-2}}{n!}.$$ 

Plugging the above into (28) with $\gamma = 1$, we obtain

$$\lim_{n \to \infty} E[L_n] = 2 + \sum_{i_1 \geq 1: \; i_2 \geq 1} \frac{S_{i_1,i_2}}{(i_1+i_2)^{i_1+i_2}}(i_1+i_2-1)!$$

$$= 2 + \sum_{n=2}^{\infty} \left( \sum_{i_1,i_2: \; i_1+i_2 = n} \frac{S_{i_1,i_2}}{n^n} \right) (n-1)!$$

$$= 2 + \sum_{n=2}^{\infty} 2 \frac{n^{n-2}}{n!} \cdot \frac{1}{n^n(n-1)!}$$

$$= 2 + \sum_{n=2}^{\infty} 2 \frac{2}{n^3} = 2\zeta(3).$$

(29)

Thus confirming the results in [15].
3.5. Proof of Theorem 3.2. Let us give a rigorous proof of Theorem 3.2. Here, we will follow the strategy used for proving Theorem 2.1 in Subsection 2.4.

Proof. Observe that
\begin{equation}
\lim_{t \to \infty} \sum_{i_1, i_2} i_1 \zeta_{i_1, i_2}(t) = 0 \quad \text{and} \quad \lim_{t \to \infty} \sum_{i_1, i_2} i_2 \zeta_{i_1, i_2}(t) = 0.
\end{equation}
Indeed, by plugging in \(\zeta_{i_1, i_2}(t)\) as in (25), we obtain
\[
\frac{d}{dt} \sum_{i_1, i_2} i_1 \zeta_{i_1, i_2}(t) = \sum_{i_1, i_2} i_1 \zeta_{i_1, i_2}(t) \left( \frac{i_1 + i_2 - 1}{t} - (\beta i_1 + \alpha i_2) \right) \leq -\alpha \wedge \beta \sum_{i_1, i_2} i_1 \zeta_{i_1, i_2}(t)
\]
for \(t > \frac{1}{\alpha \wedge \beta}\). Thus, \(\sum_{i_1, i_2} i_1 \zeta_{i_1, i_2}(t)\), and similarly \(\sum_{i_1, i_2} i_2 \zeta_{i_1, i_2}(t)\), would decrease to zero exponentially fast when \(t > \frac{1}{\alpha \wedge \beta}\).

Now, having established (30), for any given \(\epsilon \in (0, 1/4)\), we can fix \(T \gg T_{gel}\) so large that
\begin{equation}
\sum_{i_1, i_2} i_1 \zeta_{i_1, i_2}(t) \leq \frac{\alpha \epsilon}{2} \quad \text{and} \quad \sum_{i_1, i_2} i_2 \zeta_{i_1, i_2}(t) \leq \frac{\beta \epsilon}{2}.
\end{equation}
Notice that the above inequalities (31) ties \(T\) to \(\epsilon\).

Fix integers \(K_1 > 0\) and \(K_2 > 0\), and let \(R := R(K_1, K_2) = \{1, 2, \ldots, K_1\} \times \{1, 2, \ldots, K_2\}\). Since we know from Section 4 that \(\lim_{n \to \infty} \frac{\zeta_{i_1, i_2}(n)}{n} = \zeta_{i_1, i_2}(t)\) a.s. over the interval \([0, T]\) for all \(i = \begin{bmatrix} i_1 \\ i_2 \end{bmatrix} \in R\), the probability of the complement of the event
\begin{equation}
Q_{R, T, n}^\epsilon := \left\{ \sum_{i \in R} i_1 \frac{\zeta_{i_1, i_2}^{[n]}(T)}{n} \leq \frac{3}{4} \alpha \epsilon \quad \text{and} \quad \sum_{i \in R} i_2 \frac{\zeta_{i_1, i_2}^{[n]}(T)}{n} \leq \frac{3}{4} \beta \epsilon \right\}
\end{equation}
is decreasing to zero as \(n \to \infty\). Moreover,
\[
q_{R, T}(n) := P\left( Q_{R, T, n}^\epsilon \right) = O(n^{-2})
\]
by Proposition 4.3 in Subsection 4.4 as
\[
\lim_{n \to \infty} \sqrt{n} \left( \sum_{i \in R} i_1 \frac{\zeta_{i_1, i_2}^{[n]}(0)}{n} - \sum_{i \in R} i_1 \zeta_{i_1, i_2}(0) \right) = \lim_{n \to \infty} \sqrt{n} \left( \alpha [n]/n - \alpha \right) = 0
\]
and
\[
\lim_{n \to \infty} \sqrt{n} \left( \sum_{i \in R} i_2 \frac{\zeta_{i_1, i_2}^{[n]}(0)}{n} - \sum_{i \in R} i_2 \zeta_{i_1, i_2}(0) \right) = \lim_{n \to \infty} \sqrt{n} \left( \beta [n]/n - \beta \right) = 0.
\]
We know from (26) that
\[
\lim_{n \to \infty} E[L_n] = \lim_{n \to \infty} \int_0^1 E[\kappa(G(n,p))] dp - 1 = \lim_{n \to \infty} \int_0^\infty \sum_{i_1,i_2} \frac{E[\zeta_{i_1,i_2}^{[n]}(t)]}{n} e^{-t/n} dt - 1
\]
provided the latter limit exists.

We will split \( \int_0^\infty \sum_{i_1,i_2} \frac{E[\zeta_{i_1,i_2}^{[n]}(t)]}{n} e^{-t/n} dt \) as follows.

\[
\int_0^\infty \sum_{i_1,i_2} \frac{E[\zeta_{i_1,i_2}^{[n]}(t)]}{n} e^{-t/n} dt = \int_0^T \sum_{i \in R} \frac{E[\zeta_{i_1,i_2}^{[n]}(t)]}{n} e^{-t/n} dt \tag{Term I}
\]
\[
+ \int_0^T \sum_{i \not\in R} \frac{E[\zeta_{i_1,i_2}^{[n]}(t)]}{n} e^{-t/n} dt \tag{Term II}
\]
\[
+ (1 - q_{R,T}(n)) \int_0^\infty \sum_{i \in R} \frac{E[\zeta_{i_1,i_2}^{[n]}(t) \mid Q_{R,T,n}^e]}{n} e^{-t/n} dt \tag{Term III}
\]
\[
+ (1 - q_{R,T}(n)) \int_0^\infty \sum_{i \not\in R} \frac{E[\zeta_{i_1,i_2}^{[n]}(t) \mid Q_{R,T,n}^e]}{n} e^{-t/n} dt \tag{Term IV}
\]
\[
+ q_{R,T}(n) \int_T^\infty \sum_{i_1,i_2} \frac{E[\zeta_{i_1,i_2}^{[n]}(t) \mid Q_{R,T,n}^e]}{n} e^{-t/n} dt \tag{Term V}
\]

Next, we estimate the terms I-V in (33).

**Term I.** As it is proven in Section 4, \( \lim_{n \to \infty} \frac{\zeta_{i_1,i_2}^{[n]}(t)}{n} = \zeta_{i_1,i_2}(t) \text{ a.s. on } [0,T] \) for all \( i = \begin{bmatrix} i_1 \\ i_2 \end{bmatrix} \in R \). Therefore,

\[
\lim_{n \to \infty} \int_0^T \sum_{i \in R} \frac{E[\zeta_{i_1,i_2}^{[n]}(t)]}{n} e^{-t/n} dt = \sum_{i \in R} \int_0^T \zeta_{i_1,i_2}(t) dt.
\]
Term II. Observe that,

$$\sum_{i \notin R} \zeta_{i_1,i_2}(t) \leq \frac{1}{n} \sum_{i_1 > K_1} \sum_{i_2} \zeta_{i_1,i_2}(t) + \frac{1}{n} \sum_{i_1} \sum_{i_2 > K_2} \zeta_{i_1,i_2}(t)$$

$$\leq \frac{1}{K_1 n} \sum_{i_1 > K_1} i_1 \zeta_{i_1,i_2}(t) + \frac{1}{n K_2} \sum_{i_2 > K_2} i_2 \zeta_{i_1,i_2}(t)$$

$$\leq \frac{\alpha[n]}{K_1 n} + \frac{\beta[n]}{n K_2} \leq 2 \frac{\alpha}{K_1} + 2 \frac{\beta}{K_2}$$

for all $n$ large enough. Thus,

$$\int_0^T \sum_{i \notin R} \frac{E[\zeta_{i_1,i_2}(t)]}{n} e^{-t/n} dt = O \left( \frac{T}{K_1} \right) + O \left( \frac{T}{K_2} \right).$$

Term III. We define the $R$-gel to be the collection of all clusters whose mass vector is not in $R$. Let

$$M_{Rgel}(t) = \begin{bmatrix} m_1(t) \\ m_2(t) \end{bmatrix}$$

denote the total mass vector of all clusters in the $R$-gel at time $t \geq 0$.

Now, conditioning on the event $Q_{R,T,n}$, we have $m_1(t) \geq \alpha(1 - \epsilon)n$ and $m_2(t) \geq \beta(1 - \epsilon)n$ for all $t \geq T$, and $n$ large enough. Thus each cluster in $R$ will be gravitating toward the $R$-gel with the rate of at least $(\alpha \wedge \beta)(1 - \epsilon)$. Consider a cluster in $R$ at time $T$. Let $T + L$ be the time it becomes a part of the $R$-gel. Then, its contribution to the integral

$$\int_T^\infty \frac{E[\zeta_{i_1,i_2}(t) | Q_{R,T,n}]}{n} e^{-t/n} dt$$

is at most

$$\int_T^{\infty} \frac{E[1_{[T,T+L]}(t) | Q_{R,T,n}]}{n} e^{-t/n} dt \leq \frac{E[L | Q_{R,T,n}]}{n} e^{-T/n} \leq \frac{1}{(\alpha \wedge \beta)(1 - \epsilon)n}.$$

The number of clusters in $R$ at time $t \geq T$ is

$$\sum_{i \in R} \zeta_{i_1,i_2}(t) \leq \sum_{i \in R} (i_1 + i_2) \zeta_{i_1,i_2}(t) \leq (\alpha + \beta)\epsilon n.$$

Therefore,

$$\int_T^\infty \frac{E[\zeta_{i_1,i_2}(t) | Q_{R,T,n}]}{n} e^{-t/n} dt \leq \frac{(\alpha + \beta)\epsilon n}{(\alpha \wedge \beta)(1 - \epsilon)n} = \frac{2\epsilon}{1 - \epsilon} < 3\epsilon.$$
**Term IV.** We let $\mathcal{C} = \{C_1, C_2, C_3, \ldots, C_M\}$ denote the set of all clusters whose mass vectors ever exceeded $K_1$ in the first coordinate and/or ever exceeded $K_2$ in the second coordinate in the history of the process $\text{ML}_n(t)$, i.e. all clusters that were ever a part of $R$-gel. The number of clusters in $\mathcal{C}$ is less than $\alpha[n]/K_1 + \beta[n]/K_2$. For each $C_i$, the emergence time $a_i$ is the time of a merger of a pair of clusters in $R$, resulting in appearance of a new cluster $C_i$ in $R$-gel. We enumerate these clusters in the order they emerge.

Let $M_i(t) = \begin{bmatrix} m_{1,i}(t) \\ m_{2,i}(t) \end{bmatrix}$ denote the mass vector of cluster $C_i$ at time $t$. Consider a pair of clusters, $C_i$ and $C_j$, coexisting in the $R$-gel at time $t$, such that $m_{1,i}, m_{1,j} < \alpha n/2$ and $m_{2,i}, m_{2,j} < \beta n/2$. We split their merger rate into two by saying that $C_i$ absorbs $C_j$ with rate $\frac{1}{2n}(m_{1,i}(t)m_{2,j}(t) + m_{2,i}(t)m_{1,j}(t))$, and $C_j$ absorbs $C_i$ with rate $\frac{1}{2n}(m_{1,i}(t)m_{2,j}(t) + m_{2,i}(t)m_{1,j}(t))$.

There is a finite stopping time

$$t^* = \min\{t \geq 0 : \exists C_i \in \mathcal{C} \text{ with } m_{1,i}(t) \geq \alpha n/2 \text{ or } m_{2,i}(t) \geq \beta n/2\}$$

when a cluster $C_i^*$ has its mass vector satisfying either $m_{1,i^*}(t^*) \geq \alpha n/2$ or $m_{2,i^*}(t^*) \geq \beta n/2$. After time $t^*$ the rules of interactions of cluster $C_i^*$ with the other clusters in $\mathcal{C}$ change as follows. For $t > t^*$, $C_i^*$ absorbs $C_j$ with rate $\frac{1}{n}(m_{1,i^*}(t)m_{2,j}(t) + m_{2,i^*}(t)m_{1,j}(t))$, while $C_i^*$ itself cannot be absorbed by any other cluster in $\mathcal{C}$.

Let $b_i$ denote the time when cluster $C_i$ is absorbed by another cluster in collection $\mathcal{C}$. Naturally, there will be only one survivor $C_i^*$ with $b_i^* = \infty$. Let $J_i = [a_i, b_i] \cap [T, \infty)$ denote the lifespan of cluster $C_i$. Note that a cluster $C_i$ from the collection $\mathcal{C}$ existing at time $t \in [a_i, b_i)$ is absorbed into one of the clusters in the $R$-gel with the total instantaneous rate of

$$\lambda_i(t) \geq \frac{1}{2n}(m_{1,i}(t)(m_{2}(t) - m_{2,i}(t)) + m_{2,i}(t)(m_{1}(t) - m_{1,i}(t))),$$

where $m_{1}(t)$ and $m_{2}(t)$ are as defined in (34). Conditioning on the event $Q_{R,T,n}$ defined in (32), we have that if $m_{1,i}(t) < \alpha n/2$ and $m_{2,i}(t) < \beta n/2$ for $t \in J_i$, then the rate of absorption of $C_i$ into the $R$-gel is

$$\lambda_i(t) \geq \frac{1}{2n}m_{1,i}(t)\beta (1 - \epsilon)n - \frac{1}{2} + \frac{1}{2n}m_{2,i}(t)\alpha (1 - \epsilon)n - \frac{1}{2}$$

$$\geq \frac{1}{2n}m_{1,i}(t)\beta \left(\frac{3}{4}n - \frac{1}{2}\right) + \frac{1}{2n}m_{2,i}(t)\alpha \left(\frac{3}{4}n - \frac{1}{2}\right)$$

$$\geq \frac{m_{1,i}(t)\beta + m_{2,i}(t)\alpha}{8} > \frac{K_1\beta + K_2\alpha}{8}.$$

Next,

$$\int \sum_{i \notin \mathcal{R}} E_s \left[ \frac{1}{n} \sum_{i \neq i_1, i_2} Q_{R,T,n}(t) e^{-t/n} dt \right] = \int \frac{1}{n} e^{-t/n} dt + \mathcal{E}$$

(35)
where \( \int_{T}^{\infty} e^{-t/n} dt \) is due to the event \( Q_{R,T,n}^e \) which guarantees the existence of at least one component from \( C \) in the \( R\text{-gel} \) for all \( t \in [T, \infty) \) and the second term \( \mathcal{E} \) is responsible for all the times \( t \geq T \) when the number of clusters in the \( R\text{-gel} \) is greater than one. The term \( \mathcal{E} \) is bounded as follows

\[
\mathcal{E} \leq \int_{T}^{\infty} E \left[ \frac{1}{n} \sum_{i \neq i^*} 1_{J_i(t) \mid Q_{R,T,n}^e} \right] e^{-t/n} dt.
\]

Now, each cluster \( C_i \) is gravitating towards the rest of the \( R\text{-gel} \) with the rate of at least \( \frac{K_1 \beta + K_2 \alpha}{8} \). Thus, for each \( i \neq i^* \),

\[
\int_{T}^{\infty} E \left[ \frac{1}{n} J_i(t) \mid Q_{R,T,n}^e \right] e^{-t/n} dt \leq 8 \left( \frac{\alpha[n]}{K_1 \beta + K_2 \alpha} \right).
\]

Hence, since the cardinality of set \( C \) is \( M < \alpha[n] / K_1 + \beta[n] / K_2 \),

\[
\mathcal{E} < \left( \frac{\alpha[n]}{K_1} + \frac{\beta[n]}{K_2} \right) \cdot \frac{8}{n(K_1 \beta + K_2 \alpha)} = \frac{8(\alpha / K_1 + \beta / K_2)}{K_1 \beta + K_2 \alpha} + o(1),
\]

and from (35), we obtain

\[
\int_{T}^{\infty} \sum_{i \notin R} E \left[ \zeta_{i_1,i_2}^{[n]}(t) \mid Q_{R,T,n}^e \right] e^{-t/n} dt = 1 + O(K_1^{-2}) + O(K_2^{-2}) + O\left( \frac{T}{n} \right) + o(1) \quad \text{as} \quad n \to \infty.
\]

**Term V.** Here

\[
q_{R,T}(n) \int_{T}^{\infty} \sum_{i_1,i_2} E \left[ \zeta_{i_1,i_2}^{[n]}(t) \mid Q_{R,T,n}^e \right] e^{-t/n} dt \leq q_{R,T}(n) \int_{T}^{\infty} \alpha[n] + \beta[n] e^{-t/n} dt
\]

\[
\leq (\alpha[n] + \beta[n]) q_{R,T}(n) = O(n^{-1})
\]

as \( q_{R,T}(n) = O(n^{-2}) \).

Finally, by putting together the analysis in **Terms I-V** in the equation (33), we obtain for a given fixed \( \epsilon \in (0,1/4) \), sufficiently large fixed \( T \gg T_{gel} \) satisfying (31), and arbitrarily large \( K_1 \) and \( K_2 \),

\[
\int_{0}^{\infty} \sum_{i_1,i_2} E \left[ \zeta_{i_1,i_2}^{[n]}(t) \right] e^{-t/n} dt = \sum_{i \in R(K_1,K_2)} \int_{0}^{T} \zeta_{i_1,i_2}(t) dt + 1 + O\left( \frac{T}{K_1} \right) + O\left( \frac{T}{K_2} \right)
\]

\[
+ O(K_1^{-2}) + O(K_2^{-2}) + O(\epsilon) + O\left( \frac{T}{n} \right) + O(n^{-1}),
\]

(36)
which when we increase $n$ to infinity will yield
\[
\lim_{n \to \infty} \int_0^\infty \sum_{i_1,i_2} E[e^{[n]}_{i_1,i_2}(t)] \frac{1}{n} e^{-t/n} dt = \sum_{i_1,i_2} \int_0^\infty \zeta_{i_1,i_2}(t) dt + 1.
\]

4. HYDRODYNAMIC LIMITS FOR MARCUS-LUSHNIKOV PROCESSES

In [20] and [10], a certain class of Markov processes, called density dependent population process, was considered. These are jump Markov processes which depend on a certain parameter $n$ which can be interpreted depending on the context of a model. Usually it represents the population size. Many coalescent processes can be restated as a case of density dependent population processes if all cluster weights are integers. There, the total mass $n$ is the parameter representing the population size. Specifically, we may assume that the coalescent process starts with $n$ clusters of unit mass each (aka singletons). In Kurtz [20] and in Chapter 11 of Ethier and Kurtz [10], the law of large numbers and the central limit theorems were established for such density dependent population processes as $n \to \infty$. In this section we will adopt these weak limit laws for the multiplicative and cross-multiplicative coalescent processes.

4.1. Density dependent population processes. We first formulate the framework for the convergence result of Kurtz as stated in Theorem 2.1 in Chapter 11 of [10] (Theorem 8.1 in [20]). There, the density dependent population processes are defined as continuous time Markov processes with state spaces in $\mathbb{Z}^d$, and transition intensities represented as follows

\[(37) \quad q^{(n)}(k,k+\ell) = n \left[ \beta_\ell \left( \frac{k}{n} \right) + O \left( \frac{1}{n} \right) \right],\]

where $\ell, k \in \mathbb{Z}^d$, and $\beta_\ell$ is a given collection of rate functions.

In Section 5.1 of [2], Aldous observes that the results from Chapter 11 of Ethier and Kurtz [10] can be used to prove the weak convergence of a Marcus-Lushnikov process to the solutions of Smoluchowski system of equations in the case when the Marcus-Lushnikov process can be formulated as a finite dimensional density dependent population process. Specifically, the Marcus-Lushnikov processes corresponding to the multiplicative and Kingman coalescent with the monodisperse initial conditions ($n$ singletons) can be represented as finite dimensional density dependent population processes defined above.

Define $F(x) = \sum_\ell \beta_\ell(x)$. Then, Theorem 2.1 in Chapter 11 of [10] (Theorem 8.1 in [20]) states the following law of large numbers. Let $\hat{X}_n(t)$ be the Markov process with the intensities $q^{(n)}(k,k+\ell)$ given in (37), and let $X_n(t) = n^{-1} \hat{X}_n(t)$. Finally, let $|x| = \sqrt{\sum_i x_i^2}$ denote the Euclidean norm in $\mathbb{R}^d$. 
Theorem 4.1. Suppose for all compact $K \subset \mathbb{R}^d$,
\[
\sum_{\ell} |\ell| \sup_{x \in K} \beta_{\ell}(\bar{x}) < \infty,
\]
and there exists $M_K > 0$ such that
\[
|F(x) - F(y)| \leq M_K |x - y|, \quad \text{for all } x, y \in K.
\]
Suppose $\lim_{n \to \infty} X_n(0) = x_0$, and $X(t)$ satisfies
\[
X(t) = X(0) + \int_0^t F(X(s))ds,
\]
for all $T \geq 0$. Then
\[
\lim_{n \to \infty} \sup_{s \in [0,T]} |X_n(s) - X(s)| = 0 \quad \text{a.s.}
\]

4.2. Hydodynamic limit for multiplicative coalescent process. Consider a multiplicative coalescent process with kernel $K(i,j) = ij$. Recall that in the definition of a coalescent process given in Subsection 2.2, a pair of clusters with masses $i$ and $j$ coalesces at the rate $K(i,j)/n$. Consider the corresponding Marcus-Lushnikov process
\[
ML_n(t) = \left(\zeta_{1,n}(t), \zeta_{2,n}(t), \ldots, \zeta_{n,n}(t), 0, 0, \ldots\right)
\]
that keeps track for the numbers of clusters in each weight category. There, the initial conditions will be $ML_n(0) = (n, 0, 0, \ldots) = ne_1$, where $e_i$ denotes the $i$-th coordinate vector.

Next, for a fixed positive integer $K$, let $\hat{X}_n(t)$ be the restriction of process $ML_n(t)$ to the first $K$ dimensions, i.e.
\[
\hat{X}_n(t) = \left(\zeta_{1,n}(t), \zeta_{2,n}(t), \ldots, \zeta_{K,n}(t)\right)
\]
with the initial conditions $\hat{X}_n(0) = ne_1$. Apparently, $\hat{X}_n(t)$ is itself a (finite dimensional) Markov process with the following transition rates of $\hat{X}_n(t)$ stated as in (37). Let $x = (x_1, x_2, \ldots, x_K)$. Then, for any positive $1 \leq i < j \leq K$, the change vector $\ell = -e_i \pm e_j + e_i + e_j 1_{i+j \leq K}$ corresponding to a merger of clusters of respective sizes $i$ and $j$ would be assigned the rate
\[
q^{(n)}(x, x + \ell) = n ij x_i x_j = n \beta_{\ell}(x),
\]
where $\beta_{\ell}(x) = ij x_i x_j$.

For a given $1 \leq i \leq K$, the change vector $\ell = -2 e_i + e_i 1_{2i \leq K}$ corresponding to a merger of a pair of clusters of size $i$ will be assigned the rate
\[
q^{(n)}(x, x + \ell) = n \left[ \frac{i^2 x_i^2}{2} - \frac{i^2 x_i}{2n} \right] = n \left[ \beta_{\ell}(x) + O\left(\frac{1}{n}\right)\right],
\]
where $\beta_{\ell}(x) = i^2 x_i^2 / 2$. 


For a given \(1 \leq i \leq K\), the change vector \(\ell = -e_i\) corresponding to a cluster of mass \(i\) merging with a cluster of mass greater than \(K\) will be assigned the rate

\[
q^{(n)}(x, x + \ell) = nix_i \left(1 - \sum_{j=1}^{K} jx_j\right) = n\beta_\ell(x),
\]

where \(\beta_\ell(x) = ix_i \left(1 - \sum_{j=1}^{K} jx_j\right)\).

Then, by Theorem 4.1, \(X_n(t) = n^{-1}\hat{X}_n(t)\) converges to \(X(t)\) as in (40), where \(X(t)\) satisfies (39) with

\[
F(x) := \sum_{\ell} \beta_\ell(x) = \sum_{ij: 1 \leq i < j \leq K} ijx_ix_j[-e_i - e_j + e_{i+j}1_{i+j \leq K}]
\]

\[
+ \frac{1}{2} \sum_{i=1}^{K} j^2x_i^2[2e_i + e_{2i}1_{2i \leq K}] - \sum_{i=1}^{K} ix_i \left(1 - \sum_{j=1}^{K} jx_j\right) e_i
\]

\[
= \sum_{i=1}^{K} \left(-ix_i + \frac{1}{2} \sum_{1 \leq i_1, i_2 \leq K} i_1i_2x_{i_1}x_{i_2}\right) e_i.
\]

Here, \(F(x)\) is naturally satisfying the Lipschitz continuity conditions (38), and the initial conditions \(X(0) = X_n(0) = e_1\).

Observe that the system of equations (39) with \(F(x)\) as in (41) will yield the reduced system of Smoluckowski coagulation equations (5) also known as the Flory’s coagulation system [13]. Thus, for a given integer \(K > 0\) and a fixed real \(T > 0\),

\[
\lim_{n \to \infty} \sup_{s \in [0, T]} \left|n^{-1}\zeta_k(n)(s) - \zeta_k(s)\right| = 0 \quad \text{a.s.}
\]

for \(k = 1, 2, \ldots, K\).

Note that the above limit no longer requires a fixed \(K\) for each individual \(k\) in (42). However, we will mainly need the following limit in our calculations,

\[
\lim_{n \to \infty} \sup_{s \in [0, T]} \left|\sum_{k=1}^{K} n^{-1}\zeta_k,n(s) - \sum_{k=1}^{K} \zeta_k(s)\right| = 0 \quad \text{a.s.}
\]

### 4.3. Hydordynamic limit for cross-multiplicative coalescent processes.

Fix integers \(K_1 > 0\) and \(K_2 > 0\), and let \(R := R(K_1, K_2) = \{1, 2, \ldots, K_1\} \times \{1, 2, \ldots, K_2\}\). Let \(e_1\) be the standard basis vectors in \(\mathbb{R}^{K_1K_2}\), enumerated by \(i = \begin{bmatrix} i_1 \\ i_2 \end{bmatrix} \in R\). Consider
Then, for any $i$ we observe the following transition rates of $\hat{X}_{i,1,2}(t)$ to a merger of clusters of respective weights $i$ and $\zeta$ a restriction of a Marcus-Lushnikov processes with the cross-multiplicative kernel $\zeta_{i,1,2}(t)$ to $\{i_1, i_2\} \in R$. Let

$$\hat{X}_n(t) = \left\{ \hat{\zeta}_{i_1, i_2}^{[n]}(t) \right\}_{i \in R}$$

with the initial conditions $\hat{X}_n(0) = \alpha[n]e^{0'} + \beta[n]e^{0''}$, where $0' = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $0'' = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$.

We observe the following transition rates of $\hat{X}_n(t)$ stated as in (37). Let $x = \sum_{i \in R} x_ie_i$. Then, for any $i$ and $j$ in $R$, the change vector $\ell = -e_i - e_j + 1_{\{i+j \in R\}}e_{i+j}$ corresponding to a merger of clusters of respective weights $i$ and $j$ would be assigned the rate

$$q^{(n)}(x, x + \ell) = n(i_1j_2 + i_2j_1)x_ix_j = n\beta_\ell(x),$$

where $\beta_\ell(x) = (i_1j_2 + i_2j_1)x_ix_j$.

For a given $i \in R$, the change vector $\ell = -e_i$ corresponding to the merger of clusters whose weight vector is $i$ with clusters whose weight vectors are not in $R$ will be assigned the rate

$$q^{(n)}(x, x + \ell) = n\left[i_1x_1\left(\beta - \sum_{j \in R} j_2x_j\right) + i_2x_1\left(\alpha - \sum_{j \in R} j_1x_j\right)\right] = n\beta_\ell(x),$$

where $\beta_\ell(x) = i_1x_1\left(\beta - \sum_{j \in R} j_2x_j\right) + i_2x_1\left(\alpha - \sum_{j \in R} j_1x_j\right)$.

Thus, by Theorem 4.1, $X_n(t)$ converges to $X(t)$ as in (40), where $X(t)$ satisfies (39) with

$$F(x) := \sum_{\ell} \ell\beta_\ell(x) = \frac{1}{2}\sum_{i, j \in R} \left[-e_i - e_j + 1_{\{i+j \in R\}}e_{i+j}\right] (i_1j_2 + i_2j_1)x_ix_j$$

$$- \sum_{i \in R} e_1i_1x_1\left(\beta - \sum_{j \in R} j_2x_j\right) - \sum_{i \in R} e_2i_2x_1\left(\alpha - \sum_{j \in R} j_1x_j\right)$$

$$\sum_{\ell \in R} e_1\left(-\beta i_1 + \alpha i_2\right)x_1 + \frac{1}{2} \sum_{\ell, k: \ell + k = 1} \left(\ell_1k_2 + \ell_2k_1\right)x_\ell x_k$$

for a fixed $T > 0$. The system of equations (39) with $F(x)$ given in (43) will yield the reduced system of Smoluchowski coagulation equations (21). So, for a fixed a pair of positive integers $K_1$ and $K_2$, and a fixed real number $T > 0$,\n
$$\lim_{n \to \infty} \sup_{s \in [0, T]} n^{-1}\left|\hat{\zeta}_{i_1, i_2}^{[n]}(s) - \zeta_{i_1, i_2}(s)\right| = 0 \quad \text{a.s.}$$
for all \( \begin{bmatrix} i_1 \\ i_2 \end{bmatrix} \in R \). Therefore,

\[
\lim_{n \to \infty} \sup_{s \in [0,T]} \left| n^{-1} \sum_{1 \leq i_1 \leq K_1 \atop 1 \leq i_2 \leq K_2} \zeta_{i_1,i_2}^{[n]}(s) - \sum_{1 \leq i_1 \leq K_1 \atop 1 \leq i_2 \leq K_2} \zeta_{i_1,i_2}(s) \right| = 0 \quad \text{a.s.}
\]

4.4. **Central Limit Theorem and related results.** The usefulness of the framework set in [10, 20] for proving weak convergence is that the law of large numbers Theorem 4.1 is enhanced with the corresponding central limit theorem (see Theorem 4.2 below) and the large deviation theory [12]. The following central limit theorem is derived in Theorem 8.2 in [20] (and Theorem 2.3 in Chapter 11 of [10]).

**Theorem 4.2.** Suppose for all compact \( K \subset R^d \),

\[
\sum_{\ell} |\ell|^2 \sup_{x \in K} \beta_{\ell}(x) < \infty
\]

and that the \( \beta_{\ell} \) and \( \partial F \) are continuous. Suppose \( X_n \) and \( X \) are as in Theorem 4.1, and suppose \( V_n = \sqrt{n}(X_n - X) \) is such that \( \lim_{n \to \infty} V_n(0) = V(0) \), where \( V(0) \) is a constant. Then \( V_n \) converges in distribution to \( V \), which is the solution of

\[
V(t) = V(0) + U(t) + \int_0^t \partial F(X(s))V(s)ds,
\]

where \( U(t) \) is a Gaussian process and \( \partial F(X(s)) = (\partial_j F_i(X(s)))_{i,j} \).

The proof of Theorem 4.2 is based on representing \( V_n(t) \) as follows. Let \( Y_{\ell} \) be independent Poisson processes with rate one. Then,

\[
V_n(t) = V_n(0) + U_n(t) + \int_0^t \sqrt{n}(F(X_n(s)) - F(X(s)))ds,
\]

where

\[
U_n(t) = \sum_{\ell} \ell W_{\ell}^{(n)} \left( \int_0^t \beta_{\ell}(X_n(s))ds \right).
\]

\( W_{\ell}^{(n)}(u) = n^{-1/2} \hat{Y}_{\ell}(nu) \), and \( \hat{Y}_{\ell}(u) := Y_{\ell}(u) - u \) are centralized Poisson processes.

Next, we will use formula (47) in order to derive an upper bound (48) on probability \( P(|X_n(T) - X(T)| \geq \delta) \). Let us consider a simple case of a density dependent population process on \( R^d \) for which the following three conditions are satisfied.

**i:** \( V_n = \sqrt{n}(X_n - X) \) is such that \( \lim_{n \to \infty} V_n(0) = V(0) \).

**ii:** Both \( X_n(t) \) and \( X(t) \) live on a compact set \( K \).

**iii:** There are finitely many vectors \( \ell \in R^d \) such that \( \beta_{\ell}(x) > 0 \) for some \( x \in K \).
Notice that the above conditions are satisfied for the Marcus-Lushnikov processes considered here, with the general bilinear kernel as in Subsection 4.2 and with the cross-multiplicative kernel as in Subsection 4.3. Specifically, for a given \( m > 0 \), let

\[
\mathcal{K}_m = \left\{ x \in \mathbb{R}^d_+ : \sum_i x_i \leq m \right\}.
\]

Then, in Subsection 4.2, \( X_n(t), X(t) \in \mathcal{K}_2 \), and in Subsection 4.3, \( X_n(t), X(t) \in \mathcal{K}_m \), where \( m > \alpha + \beta \).

Proposition 4.3. Assuming the above conditions i-iii are satisfied together with the Lipschitz continuity conditions (38), we have

\[
P(|X_n(T) - X(T)| \geq \delta) = O(n^{-2}).
\]

Proof. Here,

\[
\sqrt{n} |F(X_n(s)) - F(X(s))| \leq \sqrt{n}M_K|X_n(s) - X(s)| = M_K|V_n(s)|
\]

and for a fixed \( T > 0 \) and any \( t \leq T \),

\[
|V_n(0) + U_n(t)| \leq \varepsilon_n(T) := |V_n(0)| + \sum_\ell |\ell| \max\left\{|W^{(n)}_{\ell}(s)| : s \in [0, T \sup_{x \in \mathcal{K}} \beta_\ell(x)]\right\}.
\]

Hence, for a fixed \( T > 0 \), equation (47) implies the following inequality,

\[
|V_n(t)| \leq \varepsilon_n(T) + M_K \int_0^t |V_n(s)|ds \quad \text{for all } t \in [0, T].
\]

Then, by Grönwall’s inequality (see Appendix 5 in [10]),

\[
|V_n(t)| \leq \varepsilon_n(T)e^{M_Kt}.
\]

In particular, we use equation (49) together with Markov inequality to obtain the following simple bound for any \( \delta > 0 \),

\[
P(|X_n(T) - X(T)| \geq \delta) \leq \frac{V_n^4(T)}{n^2 \delta^4} \leq \frac{E[\varepsilon_n^4(T)]e^{4M_KT}}{n^2 \delta^4}.
\]

Here, for any fixed real \( S > 0 \), integer \( r > 0 \), and any real \( \lambda > 0 \), we have by Doob’s martingale inequality,

\[
P\left(\max_{s \in [0, S]} |W^{(n)}_{\ell}(s)|^r \geq \lambda\right) = P\left(\max_{s \in [0, S]} |W^{(n)}_{\ell}(s)| \geq \lambda^{1/r}\right) \leq \frac{E\left[\left(W^{(n)}_{\ell}(S)\right)^{2+2r}\right]}{\lambda^{2+2/r}}.
\]
as \( |W^{(n)}_t(s)| \) is a non-negative sub-martingale. Therefore,

\[
E \left[ \max_{s \in [0,S]} \left| W^{(n)}_t(s) \right|^r \right] \leq 1 + \int_1^\infty P \left( \max_{s \in [0,S]} \left| W^{(n)}_t(s) \right|^r \geq \lambda \right) \, d\lambda \\
\leq 1 + \left( 1 + \frac{2}{r} \right) E \left[ \left( W^{(n)}_t(S) \right)^{2+2r} \right],
\]

where by the classical central limit theorem,

\[
\lim_{n \to \infty} E \left[ \left( W^{(n)}_t(S) \right)^{2+2r} \right] = S^{1+r} E[Z^{2+2r}], \quad Z - \text{standard normal r.v.}
\]

Thus, \( E[\xi_n^4(T)] = O(1) \), and (48) follows from (50). \(\square\)

5. Discussion: generalizations and open problems.

As the natural next step we see finding the limit mean length of the minimal spanning tree for random graph processes on a variety of irregular multipartite graphs via the corresponding Marcus-Lushnikov processes with multidimensional weight vectors. Note that the theory presented here extends to many other irregular graphs beyond multipartite graphs.

The coalescence dynamics of clusters with multidimensional weight (mass) vectors was originally considered in the context of aggregation kinetics [19, 31] with applications to aerosol dynamics and copolymerization kinetics. The gelation of bipartite Marcus-Lushnikov dynamics for other coalescent processes with multidimensional weight vectors is an interesting problem on its own. One may look for a generalization of the existing gelation results [29, 1, 18] in the Smoluchowski coagulation equations of the Marcus-Lushnikov processes with multidimensional weight vectors.

One of the issues facing the use of coalescent processes in genetics as models of genetic drift viewed backwards in time is that of genetic recombination. There, distinct gene loci would follow different pathways of ancestry, resulting in different gene genealogies. As a biological application, it will be useful to consider a coalescent process with multidimensional weight vectors as a means of addressing the issue of genetic recombination, and possibly, the issue of biological compatibility.

Finally, analyzing the convergence rates in the hydrodynamic limits, we could obtain a central limit theorem for \( L_n \) on \( K_{\alpha_n,\beta_n} \) similar to the central limit theorem for \( L_n \) on \( K_n \) proved in Jensen [17]. Specifically, we hope to apply Theorem 4.2 in the analysis. Also, similarly to [7], we could examine the second and third order terms in \( E[L_n] \).

References


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