A Markov process for an infinite interacting particle system in the continuum*

Yuri Kozitsky† Michael Röckner‡

Abstract

An infinite system of point particles placed in $\mathbb{R}^d$ is studied. Its constituents perform random jumps (walks) with mutual repulsion described by a translation-invariant jump kernel and interaction potential, respectively. The pure states of the system are locally finite subsets of $\mathbb{R}^d$, which can also be interpreted as locally finite Radon measures. The set of all such measures $\Gamma$ is equipped with the vague topology and the corresponding Borel $\sigma$-field. For a special class $\mathcal{P}_{\exp}$ of (sub-Poissonian) probability measures on $\Gamma$, we prove the existence of a unique family $\{P_{t,\mu} : t \geq 0, \mu \in \mathcal{P}_{\exp}\}$ of probability measures on the space of cadlag paths with values in $\Gamma$ that solves a restricted initial-value martingale problem for the mentioned system. Thereby, a Markov process with cadlag paths is specified which describes the stochastic dynamics of this particle system.

Keywords: Measure-valued Markov process; point process; martingale solution; Fokker-Planck equation; stochastic semigroup.

MSC2020 subject classifications: Primary 60J25; 60J75; 60G55, Secondary 35Q84.

Submitted to EJP on February 10, 2020, final version accepted on April 27, 2021.

Contents

1 Introduction 2

2 Preliminaries 5

2.1 The configuration spaces 5

2.2 Sub-Poissonian measures 6

2.3 Tempered configurations 8

2.4 Functions and measures on $\Gamma_*$ 12

---

*This work was supported by the Deutsche Forschungsgemeinschaft through SFB 1283 “Taming uncertainty and profiting from randomness and low regularity in analysis, stochastics and their applications” that is acknowledged by the authors.

†Uniwersytet Marii Curie-Skłodowskiej, Lublin, Poland. E-mail: jkozi@hektor.umcs.lublin.pl

‡Universität Bielefeld, Bielefeld, Germany. E-mail: roeckner@math.uni-bielefeld.de
1 Introduction

As a challenging object of probability theory, measure-valued Markov processes attract considerable attention. They have also become popular due to applications in mathematical physics, biology, ecology, etc. Among such applications one can distinguish those describing stochastic evolution of infinite systems of point particles dwelling in a continuous habitat, e.g., $\mathbb{R}^d$. In this case, as the state space of the system is taken the set of all locally finite configurations of particles, which can also be interpreted as counting Radon measures. For finite particle systems, the construction of the corresponding Markov processes is now quite standard. For infinite systems, however, the list of results reduces mostly to those describing free (noninteracting) systems [22], conservative diffusions with invariant Gibbs measures [1], or birth-and-death dynamics with generators obeying essential restrictions [17, 18, 23, 29]. In this context, one can also mention models with interactions of Curie-Weiss (mean-field) type, e.g., [26], where one starts with a system of $N$ particles interacting with a uniform strength proportional to $1/N$, and then passes to the limit $N \to +\infty$.

In the present paper, we prove the existence and uniqueness of a Markov process with cadlag paths for an infinite system of point particles performing random jumps (walks) in $\mathbb{R}^d$ with mutual repulsion, which appears to be the first result of this kind known in the literature. The starting point of our construction is the configuration space $\Gamma$. As in [25], by a configuration $\gamma$ we mean a finite or countably infinite, unordered system of points placed in $\mathbb{R}^d$, in which several points may have the same location.
A Markov process for a particle system

Configurations are supposed to be locally finite, which means that each compact \( \Lambda \subset \mathbb{R}^d \) contains a finite number of elements of a given \( \gamma \in \Gamma \). The set \( \Gamma \) is equipped with the vague (weak-hash) topology – the weakest topology that makes continuous all the maps \( \gamma \mapsto \sum_{x \in \gamma} g(x) \), \( g \in C_c(\mathbb{R}^d) \), where \( C_c(\mathbb{R}^d) \) denotes the set of all compactly supported continuous functions \( g : \mathbb{R}^d \to \mathbb{R} \). Here by writing \( \sum_{x \in \gamma} g(x) \) we understand \( \sum_i g(x_i) \) for a certain enumeration of the elements of \( \gamma \). Clearly, such sums are independent of the enumeration choice, see [25]. The vague topology is separable and consistent with a complete metric, i.e., is metrizable in such a way that the corresponding metric space is complete. Then the states of the considered system are probability measures on \( \Gamma \), the set of which is denoted by \( \mathcal{P}(\Gamma) \). The point states \( \gamma \) are associated to the Dirac measures \( \delta_\gamma \). The evolution of the system which we consider is described by the (backward) Kolmogorov equation

\[
\frac{d}{dt} F_t = LF_t, \quad (1.1)
\]

where \( F_t : \Gamma \to \mathbb{R}, t \geq 0 \), are test functions and

\[
\langle LF \rangle (\gamma) = \sum_{x \in \gamma} \int_{\mathbb{R}^d} c(x, y; \gamma) [F(\gamma \setminus x \cup y) - F(\gamma)] dy, \quad (1.2)
\]

\[
c(x, y; \gamma) = a(x - y) \exp \left( - \sum_{z \in \gamma \setminus x} \phi(z - y) \right) .
\]

Here – and in sequel in similar expressions – by writing \( \gamma \setminus x, x \in \gamma \), or \( \gamma \cup x, x \in \mathbb{R}^d \), we mean \( \gamma \setminus \{x\} \) and \( \gamma \cup \{x\} \), respectively, i.e., \( x \) is considered as the singleton \( \{x\} \).

The model specified by (1.2) presents an infinite collection of point particles performing random walks (jumps) over \( \mathbb{R}^d \), such that the remaining particles prevent the one located at \( x \in \gamma \) from jumping to \( y \) – by diminishing the jump kernel – if the target point is ‘close’ to \( \gamma \setminus x \). The diminishing factor \( \exp \left( - \sum_{z \in \gamma \setminus x} \phi(z - y) \right) \) is independent of \( x \). Originally, models of this kind were introduced and (heuristically) studied in physics [19], where they are known under a common name Kawasaki model. In the rigorous setting, the stochastic dynamics of the model described by (1.1), (1.2) were studied in [4] (see also [6] for preliminary results). In [4], for a class of states \( \mathcal{P}_{exp} \subset \mathcal{P}(\Gamma) \) – defined by a certain analytic condition – and each \( \mu_0 \in \mathcal{P}_{exp} \), there was constructed a map \( [0, +\infty) \ni t \mapsto \mu_t \in \mathcal{P}_{exp} \) that can be interpreted as the evolution of states described by (1.1). In the present work, we construct a Markov process with cadlag paths such that the mentioned \( \mu_t \) is its law at time \( t \). Let us outline now some of the aspects of this construction. As we show here, for a sufficiently large set of functions \( F : \Gamma \to \mathbb{R} \), the map \( [0, +\infty) \ni t \mapsto \mu_t \in \mathcal{P}_{exp} \) constructed in [4] is the unique (in the set of all measures) solution of the Fokker-Planck equation

\[
\mu_t(F) = \mu_0(F) + \int_s^t \mu_u(LF) du, \quad \mu(F) := \int F d\mu, \quad (1.3)
\]

holding for all \( 0 \leq s < t < \infty \), see [8] for a general theory of the equations of this kind. Unfortunately, the Dirac measure \( \delta_\gamma \) is not in \( \mathcal{P}_{exp} \) for any \( \gamma \in \Gamma \). Therefore, one cannot directly construct a transition function (and hence the corresponding Markov process) just by setting \( \mu_0 = \delta_\gamma \). In view of this, we take a version of the martingale approach suggested in [30], see also [13, Sect. 5.1], [15, Chapter 4], and proceed as follows. When dealing with measures \( \mu \in \mathcal{P}_{exp} \), it is natural to use a subset \( \Gamma_* \subset \Gamma \) such that \( \mu(\Gamma_*) = 1 \) for all \( \mu \in \mathcal{P}_{exp} \). We define it by means of a positive continuous function \( \psi : \mathbb{R}^d \to \mathbb{R} \), chosen in such a way that \( \Psi(\gamma) := \sum_{x \in \gamma} \psi(x) \) be \( \mu \)-integrable.
A Markov process for a particle system

for each \( \mu \in \mathcal{P}_{\text{exp}} \). Thereby, we set \( \Gamma_* = \{ \gamma : \Psi(\gamma) < \infty \} \), and equip it with the weakest topology that makes continuous all the maps \( \gamma \mapsto \sum_{x \in \gamma} g(x) \psi(x) \), \( g \in C_b(\mathbb{R}^d) \), where the latter is the set of all bounded continuous functions. This topology makes \( \Gamma_* \) a Polish space, continuously embedded in \( \Gamma \). Then the measures of interest are redefined as measures on \( \Gamma_* \). To construct the process in question, we use spaces of cadlag maps \( \{ s, +\infty \} \ni t \mapsto \gamma_t \in \Gamma_* \), \( s \geq 0 \), denoted by \( \mathcal{D}_{\{ s, +\infty \}}(\Gamma_*) \), equipped with the Skorohod metric, see [15, page 118], constructed with the help of a complete metric of \( \Gamma_* \). The principal result of this work (Theorem 3.6) can be characterized as follows. We prove that there exists a family of probability measures, \( \{ P_{s, \mu} : s \geq 0, \mu \in \mathcal{P}_{\text{exp}} \} \), on \( \mathcal{D}_{\mathbb{R}_+}(\Gamma_*) \) which is a unique solution of the restricted initial-value martingale problem corresponding to (1.2). For such measures, their one-dimensional marginals belong to \( \mathcal{P}_{\text{exp}} \) and satisfy the corresponding version of the Fokker-Planck equation (1.3), i.e., they coincide with the measures \( \mu_t \) constructed in [4]. By this we prove the existence of a unique Markov process with cadlag paths taking values in \( \Gamma_* \). Finally, we prove that with probability one the constructed process takes values in the subset of \( \Gamma_* \) consisting of simple configurations.

In [5], there was studied a model in which point particles of two types perform random jumps over \( \mathbb{R}^d \). Their common dynamics are described by the corresponding analog of the Kolmogorov operator (1.2) in which particles of different types repel each other, whereas those of the same type do not interact. This kind of interaction is typical for the classical Widom-Rowlinson model (see [11] and the literature quoted therein), for which the states of thermal equilibrium can be multiple [11, 24]. The latter fact ought to have an essential impact on the stochastic dynamics of such models, cf. [20], which further stimulates constructing Markov processes here. The results of [5] are pretty analogous to those of [4], which means that – after proper modification – the approach developed in the present work can be applied also to the model of [5], which we will realize in a subsequent paper.

The rest of the paper is organized as follows. In Sect. 2, we introduce all necessary facts and notions, among which are sub-Poissonian measures and the above-mentioned set \( \Gamma_* \subset \Gamma \). Here we also introduce and study two classes of functions \( F : \Gamma_* \to \mathbb{R} \), which play a crucial role in defining the Kolmogorov operator \( L \) introduced in (1.2). In Sect. 3, we impose standard assumptions on \( a \) and \( \phi \) and then make precise the domain of \( L \). Thereafter, in Theorem 3.6 we formulate the result, the main part of which is the statement that the restricted initial value martingale problem for our model has precisely one solution. Then we outline our strategy of proving this statement. In Sect. 4, we present and employ the results of [4] where the evolution of states \( t \to \mu_t \in \mathcal{P}_{\text{exp}} \) was constructed. In Sect. 5, we prove that the restricted initial value martingale problem for our model has at most one solution. This is done by proving that the Fokker-Planck equation (1.3) has a unique solution, which lies in the class of sub-Poissonian measures. Since the one-dimensional marginals of the path measures in question should solve (1.3), this yields a tool of proving the desired uniqueness. In Sect. 6 and 7, we prove the existence of the path measures by employing auxiliary models (Sect. 6) for which one can construct the processes directly (by means of transition functions), and then by proving (Sect. 7) that these models approximate the main model. Their Markov property is then obtained similarly as in [13, Sect. 5.1, pages 78, 79].

Notations and notions

In view of the size of this work, for the reader convenience we collect here essential notations and notions used throughout the whole paper.
Sets and spaces

• The habitat of the system which we study is the Euclidean space $\mathbb{R}^d$. By $\Lambda$ we always denote a compact subset of it. Further related notations: $\mathbb{R}_+ = [0, +\infty)$; $\mathbb{N} = \{1, 2, 3, \ldots\}$, $\mathbb{N}_0 = \mathbb{N} \cup \{0\}$; $C_{cs}(\mathbb{R}^d)$ - the set of all compactly supported continuous functions $g : \mathbb{R}^d \to \mathbb{R}$, $B_r(y) = \{x \in \mathbb{R}^d : |x - y| \leq r\}$, $r > 0$ and $y \in \mathbb{R}^d$. For a finite subset $\Delta \subset \mathbb{R}$, by $|\Delta|$ we denote its cardinality.

• By a Polish space we mean a separable topological space, the topology of which is consistent with a complete metric, see, e.g., [12, Chapt. 8]. Subsets of such spaces are usually denoted by $A$, $B$, whereas $A$, $B$ (with indices) are reserved for denoting operators. For a Polish space $E$, by $C_b(E)$ and $B_b(E)$ we denote the sets of bounded continuous and bounded measurable functions $g : E \to \mathbb{R}$, respectively; $B(E)$ denotes the Borel $\sigma$-field of subsets of $E$. For a suitable set $\Delta$, by $\mathcal{I}_\Delta$ we denote the indicator of $\Delta$.

• By $\Gamma, \Gamma_0, \Gamma_*$, and $\check{\Gamma}_*$, we denote configurations spaces consisting of all configurations, finite configurations (2.1), tempered configurations (2.20), and tempered simple configurations, respectively, see (2.30). These sets are equipped with the vague topology ($\Gamma$) and the weak topologies ($\Gamma_0, \Gamma_*, \check{\Gamma}_*$), which make them Polish spaces, see Lemma 2.7. By $\mathcal{P}(\Gamma), \mathcal{P}(\Gamma_*)$ we denote the sets of probability measures defined on these spaces. The set of sub-Poissonian measures $\mathcal{P}_{\text{exp}}$ is introduced in Definition 2.3. Its crucial property is established in Lemma 2.10.

• By $\mathcal{D}_{[s, +\infty)}(\Gamma_*)$ we denote the space of cadlag paths $\gamma : [0, +\infty) \to \Gamma_*$, and $\mathcal{D}_{\mathbb{R}^d}(\Gamma_*) := \mathcal{D}_{[0, +\infty)}(\Gamma_*)$. Functions on such spaces are denoted by $F, G$, etc. By $\varpi$, we denote the evaluation map, i.e., $\varpi_t(\gamma) = \gamma_t \in \Gamma_*$. Related $\sigma$-fields of measurable subsets are defined in (3.14).

Functions, measures, operators

• Functions $f : \mathbb{R}^d \to \mathbb{R}$ are usually denoted by small letters $f, g, \theta$, etc. By $\psi$ we denote the function by which we define tempered configurations, see (2.17) and (2.16). For a positive integrable $\theta : \mathbb{R}^d \to \mathbb{R}_+$, we write $\langle \theta \rangle = \int \theta(x)dx$. Functions $F : \Gamma_* \to \mathbb{R}$ are denoted by capital letters, often $F$ with additional symbols. The key functions are defined in (2.36) and (2.42). Functions defined on finite configurations $\Gamma_0$ are mostly denoted by capital $G$ with exception for correlation functions $k_{\mu,\nu}$, see (2.7).

• Measures on configuration spaces and their correlation measures are denoted by $\mu$ and $\chi_{\mu,\nu}$, respectively. Measures on $\mathbb{R}^d$ are usually denoted by $\nu$. By $\lambda$ we denote the Lebesgue-Poisson measure, see (2.3). Measures on path spaces are denoted by capital $P$. For a tempered configuration $\gamma \in \Gamma_*$, by $\nu_\gamma$ we denote the measure $\sum_{x \in \gamma} \psi(x)\delta_x$, see (2.23). The complete metric on $\Gamma_*$ used to obtain Chentsov-like estimates is defined in (2.24).

• By $L$ we denote the Kolmogorov operator (1.2), (3.1), whereas $L^\alpha$ stands for the approximating operator (6.3). Then $L^\Lambda$ and $\check{L}$ are the counterparts of $L$ acting in the spaces of functions of $\eta \in \Gamma_0$, see (4.1), (4.8) and (4.11), (4.12). By $K$ we define the operator defined in (2.4). Operators $L^{\alpha,\gamma}$ act in the Banach space of signed measures $\mathcal{M}_+$, see (7.9), (7.10).

2 Preliminaries

2.1 The configuration spaces

Each $\gamma \in \Gamma$ gives rise to a counting Radon measure $\sum_{x \in \gamma} \delta_x$. Bearing this fact in mind, we shall mostly keep using set notations, i.e., for a compact $\Lambda \subset \mathbb{R}^d$, the value of the
mentioned measure on $\Lambda$ is denoted by $|\gamma \cap \Lambda|$. The vague (weak-hash) topology of $\Gamma$ is
defined as the weakest topology that makes continuous all the maps $\Gamma \ni \gamma \mapsto \sum_{x \in \gamma} f(x)$
with $f \in C_0(\mathbb{R}^d)$. The corresponding Borel $\sigma$-field $\mathcal{B}(\Gamma)$ is the smallest $\sigma$-field of subsets
of $\Gamma$ that makes measurable all the maps $\gamma \mapsto N_\Lambda(\gamma) := |\gamma \cap \Lambda|$ with compact $\Lambda \subset \mathbb{R}^d$. By
$\mathcal{P}(\Gamma)$ we denote the set of all probability measures on $(\Gamma, \mathcal{B}(\Gamma))$.

As mentioned in Introduction, configurations $\gamma \in \Gamma$ may have multiple points. Let
$x_1, x_2, \ldots$ be any enumeration of the elements of a given $\gamma$ in which coinciding $x$ receive
distinct numbers. Then, for a suitable function $g$, by $\sum_{x \in \gamma} g(x)$ we will mean $\sum_{i} g(x_i)$,
which is independent of the enumeration used herein. The same relates to the sums
\[
\sum_{x \in \gamma} \sum_{y \in \gamma} \sum_{z \in \gamma \setminus \{x, y\}} \cdots ,
\]

Along with $\Gamma$, we also use
\[
\Gamma_0 := \bigcup_{n \in \mathbb{N}_0} \Gamma^{(n)}, \quad \Gamma^{(n)} = \{\gamma \in \Gamma : |\gamma| = n\}. \tag{2.1}
\]

Obviously, each $\Gamma^{(n)}$ - and hence the set of finite configurations $\Gamma_0$ - belong to $\mathcal{B}(\Gamma)$. The
topology induced on $\Gamma_0$ by the vague topology of $\Gamma$ coincides with the weak topology
determined with the help of $C_0(\mathbb{R}^d)$. Then the corresponding Borel $\sigma$-field $\mathcal{B}(\Gamma_0)$ is a
sub-field of $\mathcal{B}(\Gamma)$. It is possible to show that a function $G: \Gamma_0 \to \mathbb{R}$ is measurable if and
only if there exists a family of symmetric Borel functions $G^{(n)}: (\mathbb{R}^d)^n \to \mathbb{R}$, $n \in \mathbb{N}$ such that
\[
G(\{x_1, \ldots, x_n\}) = G^{(n)}(x_1, \ldots, x_n). \tag{2.2}
\]

In this context, we also write $G^{(0)} = G(\emptyset)$.

**Definition 2.1.** A measurable function, $G: \Gamma_0 \to \mathbb{R}$, is said to have bounded support if
there exist $N \in \mathbb{N}$ and a compact $\Lambda$ such that: (a) $G^{(n)} \equiv 0$ for all $n > N$; (b) $G(\eta) = 0$
whenever $\eta$ is not a subset of $\Lambda$. By $\mathcal{B}_{bs}$ we will denote the set of all bounded functions
with bounded support. For $G \in \mathcal{B}_{bs}$, $N_G$ and $\Lambda_G$ will denote the least $N$ and $\Lambda$ as in (a)
and (b), respectively. We also set $C_G = \sup_{\eta \in \Gamma_0} |G(\eta)|$.

The Lebesgue-Poisson measure $\lambda$ is defined on $\Gamma_0$ by the integrals
\[
\int_{\Gamma_0} G(\eta) \lambda(d\eta) = G(\emptyset) + \sum_{n=1}^{\infty} \frac{1}{n!} \int_{(\mathbb{R}^d)^n} G^{(n)}(x_1, \ldots, x_n) \, dx_1 \cdots dx_n, \tag{2.3}
\]

holding for all $G \in \mathcal{B}_{bs}$. For $G \in \mathcal{B}_{bs}$, we set
\[
(KG)(\gamma) := \sum_{\eta \subset \gamma} G(\eta), \quad \gamma \in \Gamma, \tag{2.4}
\]

where $\eta \subset \gamma$ means $\eta \in \Gamma_0$, i.e., the sum in (2.4) runs over finite subsets of $\gamma$.

**Remark 2.2.** [21, Proposition 3.1] For each $G \in \mathcal{B}_{bs}$, $KG$ is measurable and such that
$|(KG)(\gamma)| \leq C_G(1 + |\gamma \cap \Lambda_G|^{N_G})$ with $C_G$, $\Lambda_G$ and $N_G$ as in Definition 2.1.

### 2.2 Sub-Poissonian measures

When dealing with infinite configurations, one might expect problems (e.g., blowups)
if the dynamics start from certain $\gamma \in \Gamma$ or $\mu \in \mathcal{P}(\Gamma)$. Thus, it seems reasonable to avoid
considering such states by imposing appropriate restrictions. Another reason to do
this is gaining technical advantages, which is especially important in view of the high
complexity of the problem. The main observation here is that, for measures having finite
correlations [25], integration over $\Gamma$ can be performed in the following way
\[
\int_{\Gamma} (KG)(\gamma)\mu(d\gamma) = G(\varnothing) + \sum_{n=1}^{\infty} \frac{1}{n!} \int_{\mathbb{R}^d} G(n)(x_1, \ldots, x_n) \chi_{\mu}^{(n)}(dx_1, \ldots, dx_n),
\]
(2.5)
where $\chi_{\mu}^{(n)}$ are the correlation measures of $\mu$. That is, for a compact $\Lambda \subset \mathbb{R}^d$, $\chi_{\mu}^{(n)}(\Lambda^n)/n!$ is the $\mu$-expected value of the number of $n$-clusters of particles contained in $\Lambda$. Next, one observes that the Kolmogorov operator (1.2) contains the probability kernel $a(x-y)dy$, which is absolutely continuous with respect to Lebesgue’s measure on $\mathbb{R}^d$. In view of this, we shall demand that each $\chi_{\mu}^{(n)}$ satisfy
\[
\chi_{\mu}^{(n)}(dx_1, \ldots, dx_n) = k_{\mu}^{(n)}(x_1, \ldots, x_n) dx_1 \cdots dx_n, \quad k_{\mu}^{(n)} \in L^\infty((\mathbb{R}^d)^n), \quad n \in \mathbb{N}.
\]
(2.6)
Thereby, the right-hand side of (2.5) can be rewritten in the form, cf. (2.3),
\[
\int_{\Gamma} (KG)(\gamma)\mu(d\gamma) = \int_{\Gamma_0} k_{\mu}(\eta)G(\eta)\lambda(d\eta) =: \langle k_{\mu}, G \rangle,
\]
(2.7)
where $k_{\mu} : \Gamma_0 \to \mathbb{R}$ is defined as in (2.2). Then $k_{\mu}^{(n)}$ (resp. $k_{\mu}$) is called $n$-th order correlation function (resp. correlation function) of $\mu$. Keeping this in mind, we introduce the following class of measures. For $\theta \in C_{cs}(\mathbb{R}^d)$ and $n \in \mathbb{N}$, we write $\theta^{\otimes n}(x_1, \ldots, x_n) = \theta(x_1) \cdots \theta(x_n)$.

**Definition 2.3.** By $\mathcal{P}_{\exp}$ we denote the set of all those $\mu \in \mathcal{P}(\Gamma)$ that have finite correlations and their correlation measures satisfy
\[
\chi_{\mu}^{(n)}(\theta^{\otimes n}) \leq \kappa^n \|\theta\|_{L^1(\mathbb{R}^d)},
\]
(2.8)
holding for some $\mu$-specific $\kappa > 0$ and all $\theta \in C_{cs}(\mathbb{R}^d)$ and $n \in \mathbb{N}$.

**Remark 2.4.** It is clear from (2.8) that the map $C_{cs}(\mathbb{R}^d) \ni \theta \mapsto \chi_{\mu}^{(n)}(\theta^{\otimes n}) \in \mathbb{R}$ can be continued to a homogeneous continuous monomial of $\theta \in L^1(\mathbb{R}^d)$. One can show that $\mu \in \mathcal{P}_{\exp}$ holds if and only if each $\chi_{\mu}^{(n)}$ satisfies (2.6) with $k_{\mu}^{(n)}$ such that
\[
0 \leq k_{\mu}^{(n)}(x_1, \ldots, x_n) \leq \kappa^n,
\]
(2.9)
for the same $\kappa$ as in (2.8). Moreover, if we set
\[
F^{\theta}(\gamma) = \prod_{x \in \gamma} (1 + \theta(x)) = \exp \left( \sum_{x \in \gamma} \log (1 + \theta(x)) \right), \quad \theta \in C_{cs}(\mathbb{R}^d),
\]
(2.10)
then the map $C_{cs}(\mathbb{R}^d) \ni \theta \mapsto \mu(F^{\theta}) \in \mathbb{R}$ can be continued to a real exponential entire function of normal type of $\theta \in L^1(\mathbb{R}^d)$. The least $\kappa$ satisfying (2.9) will be called the type of $\mu$.

A Poisson measure, $\pi_{\chi}$, is characterized by its intensity measure $\chi$, see, e.g., [13, page 45], by the following formula
\[
\pi_{\chi}(F^{\theta}) = \exp (\chi(\theta)).
\]
Then $\pi_{\chi} \in \mathcal{P}_{\exp}$ if
\[
\chi(dx) = \varrho(x)dx, \quad \varrho \in L^\infty(\mathbb{R}^d).
\]
In particular, this holds for the homogeneous Poisson measure $\pi_{\varrho}$, for which $\varrho(x) \equiv \kappa > 0$. 

EJP 26 (2021), paper 72. https://www.imstat.org/ejp
A Markov process for a particle system

**Remark 2.5.** Let $G$ in (2.7) be positive, i.e., such that $G(\eta) \geq 0$ for all $\eta \in \Gamma_0$. Then by (2.9) it follows that $\mu(KG) \leq \pi_\kappa(KG)$, where $\kappa$ is the type of $\mu$. In view of this, the elements of $\mathcal{P}_\exp$ are called sub-Poissonian measures. By taking in (2.8) $\theta = I_\Lambda$ one gets that the $\mu$-expected value of the number of $n$-clusters contained in $\Lambda$ does not exceed that of the homogeneous Poissson measure with density $\kappa$, i.e., clusters are not more probable than in the case of free particles. Moreover, the states of thermal equilibrium of infinite systems of physical particles interacting via super-stable potentials belong to $\mathcal{P}_\exp$, see [27].

Recall that $I_\Lambda$ denotes the indicator of $\Lambda$. Then $N_\Lambda(\gamma) := |\gamma \cap \Lambda| = \sum_{x \in \gamma} I_\Lambda(x)$, and thus

$$
N_\Lambda^n(\gamma) = \sum_{l=1}^n S(n, l) \sum_{x_1 \in \gamma} \sum_{x_2 \in \gamma \setminus \{x_1\}} \cdots \sum_{x_l \in \gamma \setminus \{x_1, \ldots, x_{l-1}\}} I_\Lambda(x_1) \cdots I_\Lambda(x_l)
$$

$$
= \sum_{l=1}^n S(n, l) \sum_{\{x_1, \ldots, x_l\} \subset \gamma} I_\Lambda(x_1) \cdots I_\Lambda(x_l), \quad n \in \mathbb{N},
$$

where $S(n, l)$ is Stirling's number of second kind – the number of ways to divide $n$ labeled items into $l$ unlabeled groups. By (2.7) this yields

$$
\pi_\kappa(N_\Lambda^n) = \sum_{l=1}^n S(n, l) (\kappa|\Lambda|)^l = T_n(\kappa|\Lambda|), \quad n \in \mathbb{N},
$$

(2.11)

where $|\Lambda|$ is the Lebesgue measure (volume) of $\Lambda$ and $T_n$, $n \in \mathbb{N}$, are Touchard's polynomials, attributed also to J. A. Grunert, S. Ramanujan, and others, see [9, page 6]. For these polynomials, it is known that, see eq. (2.19) ibid,

$$
\exp(x(e^z - 1)) = \sum_{n=0}^{\infty} \frac{T_n(x)}{n!} z^n.
$$

(2.12)

Then for $\mu \in \mathcal{P}_\exp$, by (2.11) we obtain, cf. Remark 2.5,

$$
\mu(N_\Lambda^n) \leq T_n(\kappa|\Lambda|).
$$

(2.13)

### 2.3 Tempered configurations

When dealing with measures from $\mathcal{P}_\exp$, it might be natural to distinguish a subset $\Gamma_* \subset \Gamma$ by the condition that $\mu(\Gamma_*) = 1$ for each $\mu \in \mathcal{P}_\exp$. Obviously, the choice of such $\Gamma_*$ should also be consistent with the properties of $L$, in particular, with those of the aforementioned probability kernel $a(x - y)dy$. Let $\psi \in C_b(\mathbb{R}^d)$ be a strictly positive function that vanishes at infinity. Denote

$$
\Psi(\gamma) = \sum_{x \in \gamma} \psi(x), \quad \Gamma^\psi = \{ \gamma \in \Gamma : \Psi(\gamma) < \infty \}.
$$

(2.14)

Let $\{\psi_n\}_{n \in \mathbb{N}} \subset C_c(\mathbb{R}^d)$ be an increasing sequence such that $0 < \psi_n(x) \to \psi(x)$, $n \to +\infty$, for each $x$. Then the maps $\Gamma \ni \gamma \mapsto \Psi_n(\gamma) := \sum_{x \in \gamma} \psi_n(x)$ are vaguely continuous; hence $\{ \gamma : \Psi_n(\gamma) \leq N \}$, $N \in \mathbb{N}$ are measurable, which by (2.14) yields $\Gamma^\psi \in \mathcal{B}(\Gamma)$. Moreover, if $\tilde{\psi}$ has similar properties and satisfies $\tilde{\psi}(x) \leq \psi(x)$, $x \in \mathbb{R}^d$, then $\Gamma^\psi \subset \Gamma^{\tilde{\psi}}$. Thus, the slower the decay of $\psi$ is, the more restrictive condition is imposed on the configurations. Bearing this in mind, we will choose $\psi$ satisfying

(i) $\mu(\Psi) < \infty$; (ii) $\int_{\mathbb{R}^d} \frac{a(x)}{\psi(x)} dx < \infty$.

(2.15)

EJP 26 (2021), paper 72.

https://www.imstat.org/ejp
A Markov process for a particle system

By (2.4) and (2.7) it follows that
\[ \mu(\Psi) = \chi_\mu(\psi) = \int_{\mathbb{R}^d} k^{(1)}_\mu(x) \psi(x) dx, \]
and thus condition (i) in (2.15) turns into
\[ \langle \psi \rangle := \int_{\mathbb{R}^d} \psi(x) dx < \infty. \tag{2.16} \]

Our choice of \( \psi \) in this work is
\[ \psi(x) = \frac{1}{1 + |x|^{d+1}}, \tag{2.17} \]
which means that we prefer to be less restrictive in choosing the jump kernel \( a \) at the expense of stronger restrictions imposed on the configurations.

Similarly as in (2.13), for all \( n \in \mathbb{N} \) and each \( \mu \in \mathcal{P}_{\text{exp}} \), one obtains
\[ \mu(\Psi^n) \leq \sum_{l=1}^n S(n, l) (\chi(\psi))^l = T_n (\chi(\psi)), \tag{2.18} \]
where we have taken into account that \( \psi^n(x) \leq \psi(x) \) for all \( n \geq 1 \) and \( x \), \( \chi \) is the type of \( \mu \). By (2.18) and (2.12) it follows that
\[ \int_{\Gamma} \exp (\beta \Psi(\gamma)) \mu(d\gamma) \leq \exp (\chi(\psi)(e^\beta - 1)), \tag{2.19} \]
holding for all \( \beta > 0 \). Next, we define
\[ \Gamma_* = \Gamma^\psi, \tag{2.20} \]
with \( \psi \) as in (2.17). By (2.18) it follows that
\[ \forall \mu \in \mathcal{P}_{\text{exp}} \quad \mu(\Gamma_*) = 1. \tag{2.21} \]

This crucial property of the elements of \( \mathcal{P}_{\text{exp}} \) will allow us to consider only configurations belonging to \( \Gamma_* \). In particular, this means that we will use the following sub-field of \( \mathcal{B}(\Gamma) \):
\[ A_* = \{ A \in \mathcal{B}(\Gamma) : A \subset \Gamma_* \}. \tag{2.22} \]

Now let us consider
\[ C^L_\nu(\mathbb{R}^d) = \{ g \in C_\nu(\mathbb{R}^d) : \| g \|_L < \infty \}, \quad \| g \|_L := \sup_{x,y \in \mathbb{R}^d, x \neq y} \frac{|g(x) - g(y)|}{|x - y|}, \]
and then define
\[ \| g \|_{BL} = \| g \|_L + \sup_{x \in \mathbb{R}^d} |g(x)|, \quad g \in C^L_\nu(\mathbb{R}^d), \]
and also
\[ \nu(\nu, \nu') = \sup_{g : \| g \|_{BL} \leq 1} |\nu(g) - \nu'(g)|, \quad \nu, \nu' \in \mathcal{N}, \]
where \( \mathcal{N} \) is the set of all positive finite measures Borel on \( \mathbb{R}^d \).

**Proposition 2.6.** [14, Theorem 18] The following three types of the convergence of a sequence \( \{ \nu_n \} \subset \mathcal{N} \) to a certain \( \nu \in \mathcal{N} \) are equivalent:

(i) \( \nu_n(g) \to \nu(g) \) for all \( g \in C_\nu(\mathbb{R}^d) \);
(ii) \( \nu_n(g) \to \nu(g) \) for all \( g \in C^L_\nu(\mathbb{R}^d) \);
A Markov process for a particle system

(iii) \( v(\nu_n, \nu) \to 0 \).

By means of this statement we prove the following important facts. For a configuration, \( \gamma \in \Gamma_* \), by \( \nu_\gamma \in \mathcal{N} \) we mean the measure defined by

\[
\nu_\gamma(g) = \sum_{x \in \gamma} g(x) \psi(x), \quad g \in C_b(\mathbb{R}^d).
\]  

(2.23)

Then we set

\[
v_*(\gamma, \gamma') = v(\nu_\gamma, \nu_{\gamma'}) = \sup_{g : \|g\|_{\infty} \leq 1} \left| \sum_{x \in \gamma} g(x) \psi(x) - \sum_{x \in \gamma'} g(x) \psi(x) \right|, \quad \gamma, \gamma' \in \Gamma_*.
\]  

(2.24)

In the next statement, by \( \hat{\Gamma}_* \) we mean the subset of \( \Gamma_* \) consisting of single configurations. That is, \( \gamma \in \Gamma_* \) belongs to \( \hat{\Gamma}_* \) if \( B_{\delta}(x) \cap \gamma = \{x\} \), holding for each \( x \in \gamma \) and an \( x \)-specific \( \delta > 0 \).

**Lemma 2.7.** The metric space \((\Gamma_*, \nu_*)\) is complete and separable. \( \hat{\Gamma}_* \) is a \( G_\delta \) subset of \( \Gamma_* \), and thus is a Polish spaces.

**Proof.** First, we prove that \( \Gamma_* \) has the properties in question. Let \( \{\gamma_n\}_{n \in \mathbb{N}} \subset \Gamma_* \) be a \( \nu_* \)-Cauchy sequence. Since the metric space \((\mathcal{N}, v)\) is complete, see [7, Corollary 8.6.3, Sect. 8.6], the sequence \( \{\nu_{\gamma_n}\}_{n \in \mathbb{N}} \) converges to a certain \( \nu \in \mathcal{N} \). As each \( h \in C_{cs}(\mathbb{R}^d) \) can be written in the form \( h(x) = g(x) \psi(x), g \in C_{cs}(\mathbb{R}^d) \), this convergence implies the vague convergence of \( \{\gamma_n\}_{n \in \mathbb{N}} \) to a certain \( \gamma \in \Gamma \). Let now \( \{g_m\}_{m \in \mathbb{N}} \subset C_{cs}(\mathbb{R}^d) \) be such that \( g_m(x) = 1 \) for \( |x| \leq m \) and \( g_m(x) = 0 \) for \( |x| \geq m + 1 \), which is possible by Urysohn’s lemma. Then

\[
\lim_{n \to +\infty} \sum_{x \in \gamma_n} g_m(x) \psi(x) = \sum_{x \in \gamma} g_m(x) \psi(x) \leq \nu(\mathbb{R}^d),
\]

which by the dominated convergence theorem yields \( \gamma \in \Gamma_* \), and hence \( \nu = \nu_\gamma \). Then \((\Gamma_*, \nu_*)\) is a complete metric space. Its separability follows by the separability of \( \mathbb{R}^d \).

When dealing with a topological property of a subset of \( \Gamma_* \), we may use any metric consistent with its weak topology. As such one, we take Prohorov’s metric, cf. [15, page 96], introduced as follows. For \( \varepsilon > 0 \) and \( \Delta \subset \mathbb{R}^d \), we set \( \Delta^\varepsilon = \cup_{x \in \Delta} B_\varepsilon(x) \) and also

\[
v_\varepsilon(\gamma, \gamma') = \inf \{\varepsilon > 0 : \nu_\gamma(\Delta^\varepsilon) \leq \nu_{\gamma'}(\Delta^\varepsilon) + \varepsilon, \ \& \ \nu_{\gamma'}(\Delta^\varepsilon) \leq \nu_\gamma(\Delta^\varepsilon) + \varepsilon, \ \forall \Delta - \text{closed}\}.
\]  

(2.25)

Let \( \{R_k\}_{k \in \mathbb{N}} \) be such that \( 0 < R_1 < R_2 < \cdots < R_k < \cdots \) and \( \lim_{k \to +\infty} R_k = +\infty \). Set \( D_k = \{x \in \mathbb{R}^d : |x| < R_k\} \) and \( \gamma_k = \gamma \cap D_k, \ \gamma \in \Gamma_* \). \( k \in \mathbb{N} \). By (2.17) we then have

\[
\sup_{x \in D_k} 1/\psi(x) = 1 + R_k^{d+1} =: \alpha_k^{-1},
\]  

(2.26)

\[
|\psi(x) - \psi(y)| \leq (d + 1)|x - y|, \quad x, y \in D_k.
\]

Next, we set

\[
\Gamma_{\ast,k} = \{\gamma \in \Gamma_* : \gamma_k \in \hat{\Gamma}_*\}, \quad k \in \mathbb{N}.
\]  

(2.27)

i.e., \( \gamma \in \Gamma_* \) belongs to \( \Gamma_{\ast,k} \) if its part in \( D_k \) is a simple configuration. Our aim is to show that \( \Gamma_{\ast,k} \) is an open subset of the Polish space \( \Gamma_* \). To this end, we take any \( \gamma \in \Gamma_{\ast,k} \) and look for \( r > 0 \) such that

\[
W_r(\gamma) := \{\gamma' : v_\varepsilon(\gamma, \gamma') < r\} \subset \Gamma_{\ast,k}.
\]

For the chosen \( \gamma \), we pick \( \ell > 0 \) satisfying \( B_\ell(x) \cap \gamma = \{x\} \) for all \( x \in \gamma_k \). Now take positive \( \varepsilon \) and \( \delta \) such that

\[
\varepsilon \leq \frac{1}{4} \min\{\ell; \alpha_k\}, \quad \delta \leq \frac{1}{4} \min\{\ell; \alpha_k/(d + 1)\},
\]  

(2.28)
A Markov process for a particle system

and then assume that $\gamma' \in \Upsilon_{r}(\gamma)$ with $r < \varepsilon$. For $x \in \gamma_k$, the second estimate in (2.25) for $\Delta = B_d(x)$ yields in this case

$$\sum_{y \in \gamma \cap D_k(x)} \psi(y) \leq \psi(x) + \varepsilon,$$

(2.29)

where we have taken into account that $B_d(x) \subseteq B_t(x)$, see (2.28). For $y \in B_d(x)$, by (2.26) we have $\psi(y) \geq \psi(x) - \delta(d + 1)$. Thus,

$$\text{LHS}(2.29) \geq m(x)\psi(x) - m(x)\delta(d + 1),$$

where $m(x) = |\gamma' \cap B_d(x)|$. Then

$$m(x) - 1 \leq \frac{\varepsilon}{\psi(x)} + \frac{m(x)\delta(d + 1)}{\psi(x)} \leq (m(x) + 1)/4,$$

which means that $m(x) = 1$, holding for each $x \in \gamma_k$ and $B_d(x)$. At the same time, for $\gamma' \in \Upsilon_{r}(\gamma)$ with $r < \varepsilon$, it follows that $\gamma' \cap (D_k \setminus \bigcup x \in \gamma_k B_d(x)) = \emptyset$. For otherwise, the second estimate in (2.25) with $\Delta = B_d(y)$, $y$ lying in the mentioned intersection, would yield $\psi(y) \leq \varepsilon$ which contradicts (2.28). Thus, $\gamma' \in \Gamma_{*\mu}$, and hence the latter is an open subset of $\Gamma_*$. Therefore, $\Gamma_* \cap k \in \mathbb{N} \Gamma_{*,k}$ is a $G_\delta$-subset of $\Gamma_*$. In view of the first part of this statement, $\Gamma_*$ is a Polish space, see [12, Proposition 8.1.5, page 242]. This completes the proof.

The following formulas summarize the relationships between the configuration spaces we will deal with

$$\hat{\Gamma}_* \subseteq \Gamma_* \subseteq \Gamma.$$  

(2.30)

Note that the embedding of the Polish space $\Gamma_*$ into the Polish space $\Gamma$ is continuous, since the weak convergence $\gamma_n \to \gamma$ implies also the corresponding vague convergence. Let $B(\Gamma_*)$, $\hat{B}(\Gamma_*)$ be the Borel $\sigma$-field of subsets of $\Gamma_*$ and $\hat{\Gamma}_*$, respectively. Recall that we have another $\sigma$-field, $\mathcal{A}_*$, defined in (2.22).

**Corollary 2.8.** It follows that $\mathcal{A}_* = B(\Gamma_*)$, $\{A \in B(\Gamma_*) : A \subset \Gamma_* \} = B(\hat{\Gamma}_*)$.

**Proof.** The first equality follows by the continuity of the embedding and then by Kuratowski’s theorem, see [28, Theorem 3.9, page 21]. The second equality follows by the equality of the weak topology of $\hat{\Gamma}_*$ with that induced by the weak topology of $\Gamma_*$. □

**Remark 2.9.** The latter statement allows one to redefine each $\mu \in \mathcal{P}(\Gamma)$ with the property $\mu(\Gamma_*) = 1$ as a measure on the measurable space $(\Gamma_*, B(\Gamma_*))$. And similarly, each measure on $(\Gamma_*, B(\Gamma_*))$ possessing the property $\mu(\Gamma_*) = 1$ can be considered as a measure on $(\hat{\Gamma}_*, B(\hat{\Gamma}_*))$.

Now we turn to proving the following statement.

**Lemma 2.10.** For each $\mu \in \mathcal{P}_{\text{exp}}$, see Definition 2.3, it follows that $\mu(\hat{\Gamma}_*) = 1$. Hence, this $\mu$ can be redefined as a measure on $(\hat{\Gamma}_*, B(\hat{\Gamma}_*))$, cf. Corollary 2.8 and Remark 2.9.

**Proof.** By our assumption the correlation measures $\lambda^{(n)}_\mu$ of the measure under consideration have the properties corresponding to (2.6) and (2.9). For $N \in \mathbb{N}$ and $\epsilon \in (0, 1)$, we set

$$H_N(\gamma) = \sum_{x \in \gamma} \sum_{y \in \gamma \setminus x} h_N(x, y), \quad h_N(x, y) = \psi(x)\psi(y) \min\{N; |x - y|^{-d\epsilon}\}.$$
Note that $H_N(\gamma) < \infty$ for all $N \in \mathbb{N}$ and $\gamma \in \Gamma$. By (2.7) we then have
\[
\mu(H_N) = \int_{(\mathbb{R}^d)^2} k^2_{x,y}(x,y)h_N(x,y)dx\,dy \leq \lambda^2 \int_{\mathbb{R}^d} \psi(x) \left( \int_{\mathbb{R}^d} \psi(y)\,dy \right) dx
\]
\[
\leq \lambda^2 \int_{\mathbb{R}^d} \psi(x) \left( \int_{B_{\lambda}(x)} \frac{dy}{|x-y|^{d\varepsilon}} + \frac{\langle \psi \rangle}{\varepsilon} \right) dx =: \lambda^2 C,
\]
for an appropriate $C > 0$. Since $H_N \leq H_{N+1}$, we can apply here the Beppo Levi (monotone convergence) theorem, which yields that the point-wise limit
\[
\lim_{N \to +\infty} H_N(\gamma) = H(\gamma) := \sum_{x \in \gamma} \sum_{y \in \gamma(x)} \psi(x)\psi(y) \frac{1}{|x-y|^{d\varepsilon}}
\]
(2.31)
is finite for $\mu$-almost all $\gamma$, i.e., for all $\gamma \in \Gamma_{*,\mu}$ such that $\mu(\Gamma_{*,\mu}) = 1$. For $c > 0$, we set $\Gamma_c = \{ \gamma : H(\gamma) \leq c \}$. Then $|x-y| \geq e^{-1/d\varepsilon}$ for all pairs $x,y \in \gamma$ and each $\gamma \in \Gamma_c$. That is, $\gamma$ is simple; hence, $\Gamma_{*,\mu} \subset \Gamma_{\epsilon,*}$, which completes the proof.

**Remark 2.11.** By (2.21) it follows that the class of measures $\mu \in \mathcal{P}(\Gamma_\epsilon)$ with the property $\mu(\Gamma_{\epsilon,*}) = 1$ includes $\mathcal{P}_{\text{exp}}$. Therefore, depending on the context, we can and will consider such measures on either of these spaces.

### 2.4 Functions and measures on $\Gamma_\epsilon$

The main aim of this part is to introduce suitable classes of functions $F : \Gamma_\epsilon \to \mathbb{R}$, for which we define $LF$ and then use in (1.3). We begin by introducing suitable functions $g : \mathbb{R}^d \to \mathbb{R}$. For $\psi$ defined in (2.17), we set
\[
\Theta_\psi = \{ \theta(x) = g(x)\psi(x) : g \in C_b(\mathbb{R}^d), \ \theta(x) \geq 0 \},
\]
\[
\Theta_\psi^+ = \{ \theta \in \Theta_\psi : \theta(x) > 0 \ \forall x \in \mathbb{R}^d \}.
\]
Clearly, each $\theta \in \Theta_\psi$ is integrable. For such $\theta$, we also define
\[
c_\theta = \sup_{x \in \mathbb{R}^d} \frac{1}{\psi(x)} \log (1 + \theta(x)), \quad c_\theta := e^{c_\theta} - 1.
\]
(2.33)
Then
\[
0 \leq \theta(x) \leq c_\theta \psi(x), \quad \theta \in \Theta_\psi.
\]
(2.34)
Now let us turn to $F^\theta$ defined in (2.10). By Remark 2.5, (2.16), Remark 2.11, and then by (2.34), for $\mu \in \mathcal{P}_{\text{exp}}$ of type $\varkappa$ we have
\[
\int_{\Gamma_{\chi}} F^\theta(\gamma)\mu(d\gamma) = \int_{\Gamma_{\chi}} F^\theta(\gamma)\mu(d\gamma) \leq \pi_{\varkappa}(F^\theta) \leq \exp (\varkappa c_\theta \psi)\mu(d\gamma), \quad \theta \in \Theta_\psi.
\]

**Remark 2.12.** In general, for $\theta \in \Theta_\psi$ the map $\Gamma_{\chi} \ni \gamma \mapsto \sum_{x \in \gamma} \theta(x)$ need not be vaguely continuous. But it is weakly continuous for all such $\theta$, which is also the case for $\Gamma_{\chi} \ni \gamma \mapsto \sum_{x \in \gamma} \theta(x)$. In particular, the map $\Gamma_{\chi} \ni \gamma \mapsto \psi(\gamma)$ is weakly continuous, that is one of the advantages of passing to tempered configurations. Since the measurability and continuity of $F : \Gamma_{\chi} \to \mathbb{R}$ and $F : \Gamma_{\chi} \to \mathbb{R}$ occur simultaneously, each such a function can and will be considered as a map acting from either of these spaces. In the sequel, when we speak of the properties of a given $F : \Gamma_{\chi} \to \mathbb{R}$, we tacitly assume that the same also holds for its restriction to $\Gamma_{\epsilon,*}$.
Then assuming the latter we conclude that

\[ g \text{To see this, for a given view of (2.38). By (2.40) and (2.39), for all } 26 \in EJP, \text{ we have that } \theta + \theta' + \theta\theta' \text{ belongs to } \Theta_\psi \text{ for each } \theta, \theta' \in \Theta_\psi. \] Next, define

\[ \tilde{F}^\theta_{\tau}(\gamma) = \prod_{x \in \gamma} (1 + \theta(x)) e^{-\tau \psi(x)} = \exp \left( - \nu_\gamma (\nu_\gamma) \right). \] (2.36)

Recall here that \( \tau > c_0 \), see (2.35). We extend this to \( \tau = 0 \) and \( \theta(x) \equiv 0 \) by setting \( \tilde{F}^0_0(\gamma) \equiv 1 \) and include this function in the set

\[ \tilde{F} := \{ \tilde{F}^\theta_{\tau} : \theta \in \Theta_\psi, \tau > c_0 \} \subset C_b(\Gamma_\star). \] (2.37)

Similarly as in [13, Sect. 3.2, page 41], see also [15, page 111], we introduce the following notion.

**Definition 2.13.** A sequence of bounded measurable functions \( F_n : \Gamma_\star \to \mathbb{R}, n \in \mathbb{N} \) is said to be bounded and pointwise (bp) convergent to a given \( F : \Gamma_\star \to \mathbb{R} \) if: (a) \( F_n(\gamma) \to F(\gamma) \) for all \( \gamma \in \Gamma_\star; \) (b) \( \sup_{n \in \mathbb{N}} \sup_{\gamma \in \Gamma} |F_n(\gamma)| < \infty. \) The \( bp \)-closure of a set \( \mathcal{H} \subset C_b(\Gamma_\star) \) is the smallest subset of \( B_b(\Gamma_\star) \) that contains \( \mathcal{H} \) and is closed under the \( bp \)-convergence. In a similar way, one defines also the bp-convergence of sequences of functions \( g : \mathbb{R}^d \to \mathbb{R}. \)

It is well-known that \( C_b(\mathbb{R}^d) \) contains a countable family of nonnegative functions, \( \{ g_i \}_{i \in \mathbb{N}}, \) which is convergence determining and such that its linear span is bp-dense in \( B_b(\mathbb{R}^d), \) see [15, Proposition 4.2, page 111] and [13, Lemma 3.2.1, page 41]. This means that a sequence of finite positive measures \( \{ \nu_n \} \in \mathcal{P}(\mathbb{R}^d) \) weakly converges to a certain \( \nu \) if and only if \( \nu_n(g_i) \to \nu(g_i), \) \( n \to +\infty \) for all \( i \in \mathbb{N}. \) One may take such a family containing the constant function \( g(x) \equiv 1 \) and closed with respect to the pointwise addition. Moreover, one may assume that

\[ \forall i \in \mathbb{N} \inf_{x \in \mathbb{R}^d} g_i(x) =: c_i > 0. \] (2.38)

If this is not the case for a given \( g_i, \) in place of it one may take \( \tilde{g}_i(x) = g_i(x) + q_i, \) with some \( q_i > 0. \) The new set, \( \{ \tilde{g}_i \}, \) has both mentioned properties and also satisfies (2.38). Then assuming the latter we conclude that

\[ V_0 := \{ g_i \}_{i \in \mathbb{N}} \subset V. \] (2.39)

To see this, for a given \( g_i, \) take \( \tau_i \geq \sup_x g_i(x) \) and then set

\[ \theta_i(x) = \exp \left( \frac{[\tau_i - g_i(x)]\psi(x)}{\tau_i} - 1. \right) \] (2.40)

Clearly, \( \theta_i(x) \geq 0. \) Since \( \psi^n(x) \leq \psi(x), \) \( n \in \mathbb{N}, \) we have that \( \theta_i(x) \leq e^{\tau_i} \psi(x), \) and hence \( \{ \theta_i \}_{i \in \mathbb{N}} \subset \Theta_\psi, \) see (2.32). At the same time, \( v_{\tau_i}^{\theta_i} = g_i \) and \( c_0 = \sup_x (\tau_i - g_i(x)) < \tau_i \) in view of (2.38). By (2.40) and (2.39), for all \( i \in \mathbb{N}, \) it follows that

\[ F_i \in \tilde{F}, \quad F_i(\gamma) := \exp \left( -\nu_\gamma (\gamma) \right). \]

**Proposition 2.14.** The set \( \tilde{F} \) defined in (2.37) is closed with respect to the pointwise multiplication. Moreover, it has the following properties:

(i) It is separating: \( \mu_1(F) = \mu_2(F), \) holding for all \( F \in \tilde{F}, \) implies \( \mu_1 = \mu_2 \) for all \( \mu_1, \mu_2 \in \mathcal{P}(\Gamma_\star). \)
A Markov process for a particle system

(ii) It is convergence determining: if a sequence \( \{ \mu_n \}_{n \in \mathbb{N}} \subset \mathcal{P}(\Gamma_+) \) is such that \( \mu_n(F) \to \mu(F) \), \( n \to +\infty \) for all \( F \in \tilde{F} \) and some \( \mu \in \mathcal{P}(\Gamma_+) \), then \( \mu_n(F) \to \mu(F) \) for all \( F \in C_b(\Gamma_+) \).

(iii) The set \( B_1(\Gamma_+) \) is the \( b \)-closure of the linear span of \( \tilde{F} \).

**Proof.** The closedness of \( \tilde{F} \) under multiplication follows directly by (2.36) and the fact that \( \theta_1 + \theta_2 + \theta_1 \theta_2 \in \Theta_\psi \) for each \( \theta_1, \theta_2 \in \Theta_\psi \). It is clear that \( \tilde{F} \) separates points of \( \Gamma_+ \), i.e., one finds \( F \in \tilde{F} \) such that \( F(\gamma_1) \neq F(\gamma_2) \) whenever \( \gamma_1 \neq \gamma_2 \), that holds for each pair \( \gamma_1, \gamma_2 \in \Gamma_+ \). Then claim (i) follows by [15, claim (a) Theorem 4.5, page 113]. Claim (ii) follows by the fact that \( \{ F_i \}_{i \in \mathbb{N}} \subset \tilde{F} \) has the property in question, which in turn follows by [13, Theorem 3.2.6, page 43]. Likewise, claim (iii) follows by [13, Lemma 3.2.5, page 43].

\( \square \)

Note that each function as in (2.36) can be written in the form

\[
\tilde{F}_\tau^\varphi(\gamma) = \exp(-\tau \Psi(\gamma)) F_\varphi(\gamma),
\]

(2.41)

where \( F_\varphi \) is as in (2.10), which is a \( \psi_\gamma \)-continuous function for each \( \theta \in \Theta_\psi \).

For \( m \in \mathbb{N} \), \( \theta_1, \ldots, \theta_m \in \Theta_\psi^+ \), see (2.32), we set

\[
\tilde{F}_\tau^{\theta_1, \ldots, \theta_m}(\gamma) = \sum_{x_1 \in \gamma} \theta_1(x_1) \sum_{x_2 \in \gamma \setminus x_1} \theta_2(x_2) \cdots \sum_{x_m \in \gamma \setminus \{x_1, \ldots, x_{m-1}\}} \theta_m(x_m) \tilde{F}_\tau^\varphi(\gamma \setminus \{x_1, \ldots, x_m\})
\]

(2.42)

where \( S_m \) is the symmetric group and \( \tilde{F}_\tau^\varphi(\gamma) = \exp(-\tau \Psi(\gamma)) \), see (2.41).

**Proposition 2.15.** For each \( \tau > 0 \), \( m \in \mathbb{N} \) and \( \theta_1, \ldots, \theta_m \in \Theta_\psi^+ \), it follows that \( \tilde{F}_\tau^{\theta_1, \ldots, \theta_m} \in C_b(\Gamma_+) \).

**Proof.** To prove the continuity of \( \tilde{F}_\tau^{\theta_1, \ldots, \theta_m} \) we rewrite (2.42) in the form

\[
\tilde{F}_\tau^{\theta_1, \ldots, \theta_m}(\gamma) = \exp(-\tau \Psi_0(\gamma)) \times \sum_{\{x_1, \ldots, x_m\} \subset \gamma} \varphi_{\sigma(1)}(x_1) \cdots \varphi_{\sigma(m)}(x_m),
\]

(2.43)

with \( \varphi_j(x) := \theta_j(x)c_{\psi}(x) \), \( j = 1, \ldots, m \). Clearly, all \( \varphi_j \) belong to \( \Theta_\psi^+ \). By an inclusion-exclusion formula the right-hand side of (2.43) can be written as a linear combination of the products of the following terms

\[
\Phi_{i_1, i_2, \ldots, i_n}(\gamma) = \sum_{x \in \gamma} \varphi_{i_1}(x) \cdots \varphi_{i_n}(x),
\]

multiplied by a continuous function, \( \gamma \mapsto \exp(-\tau \Psi(\gamma)) \). Since \( \Theta_\psi^+ \) is closed with respect to the pointwise multiplication, such terms are continuous that yields the continuity of \( \tilde{F}_\tau^{\theta_1, \ldots, \theta_m} \). To prove the boundedness we estimate each \( \varphi_j(x) \leq \varphi(x) := c_{\psi}(x)e^{\tau \Psi(x)} \leq c_{\psi}(x)e^{\tau} \). Then

\[
\tilde{F}_\tau^{\theta_1, \ldots, \theta_m}(\gamma) \leq \exp(-\tau \Psi(\gamma)) \sum_{x \in \gamma} \varphi(x_1) \sum_{x_2 \in \gamma \setminus x_1} \varphi(x_2) \cdots \sum_{x_m \in \gamma \setminus \{x_1, \ldots, x_{m-1}\}} \varphi(x_m)
\]

\[
\leq \exp(-\tau \Psi(\gamma)) \left( \sum_{x \in \gamma} \varphi(x) \right)^m \leq c_{\psi}^m \Psi^m(\gamma) \exp(-\tau [\Psi(\gamma) - m])
\]

\[
\leq \left( \frac{cm}{\tau} \right)^m \exp(m(\tau - 1)),
\]

EJP 26 (2021), paper 72.
which completes the proof.

3 The Result

3.1 The domain of $L$

Here we recall that the model we study is specified by the Kolmogorov operator, cf. (1.2),

$$(LF)(\gamma) = \sum_{x \in \gamma} \int_{\mathbb{R}^d} a(x-y) \exp \left( - \sum_{z \in \gamma \setminus x} \phi(z-y) \right) \left[ F(\gamma \setminus x \cup y) - F(\gamma) \right] dy.$$  

(3.1)

Let us make precise the conditions imposed on the model. The positive measurable functions $a$ and $\phi$ are supposed to satisfy the following:

$$\sup_x a(x) = a < \infty, \quad \sup_x \phi(x) = \bar{\phi} < \infty,$$

(3.2)

$$\int_{\mathbb{R}^d} \phi(x) dx =: \langle \phi \rangle < \infty, \quad \int_{\mathbb{R}^d} a(x) dx = 1,$$

and

$$\int_{\mathbb{R}^d} |x|^l a(x) dx =: m_l^a < \infty, \quad \text{for } l = 1, \ldots, d + 1.$$  

(3.3)

The conditions in (3.2) are the same as in [4]. We impose them to be able to use the results of this work here. Note that the assumed boundedness of $\phi$ excludes a hard-core repulsion. The condition in (3.3) is the realization of item (ii) of (2.15). It was not used in [4].

As mentioned in Introduction, we are going to construct the process as a solution of a restricted initial value martingale problem. In this case, the domain of the operator introduced in (1.2) is crucial, cf. [13, page 79]. Along with the set introduced in (2.37), we define

$$\hat{F} = \{ \hat{F}_{\theta_1, \ldots, \theta_m} : m \in \mathbb{N}, \theta_1, \ldots, \theta_m \in \Theta^+_\psi, \tau > 0 \},$$

(3.4)

where $\hat{F}_{\theta_1, \ldots, \theta_m}$ is as in (2.42).

Definition 3.1. By $\mathcal{D}(L)$ we denote the linear span of the set $\hat{F} \cup \hat{F}$.

By (2.37) and Proposition 2.15 one concludes that $\mathcal{D}(L) \subset C_0(\Gamma_*)$. Let us show that $L\hat{F}_{\theta_1, \ldots, \theta_m} \in B_0(\Gamma_*)$. For $\gamma \in \Gamma_*, x \in \gamma, y \in \mathbb{R}^d$ and a suitable $F : \Gamma_* \to \mathbb{R}$, define, cf. (1.2),

$$\nabla^{y.x} F(\gamma) = F(\gamma \setminus x \cup y) - F(\gamma).$$

By (2.42) we have

$$\hat{F}_{\theta_1, \ldots, \theta_m}(\gamma) = \sum_{x_1 \in \gamma} \theta_1(x_1) \hat{F}_{\theta_2, \ldots, \theta_m}(\gamma \setminus x_1).$$

Then

$$\nabla^{y.x} \hat{F}_{\theta_1, \ldots, \theta_m}(\gamma) = \theta_1(y) - \theta_1(x_1) \hat{F}_{\theta_2, \ldots, \theta_m}(\gamma \setminus x_1) + \sum_{x_1 \in \gamma \setminus x} \theta_1(x_1) \nabla^{y.x} \hat{F}_{\theta_2, \ldots, \theta_m}(\gamma \setminus x_1).$$

By iterating the latter we get

$$\nabla^{y.x} \hat{F}_{\theta_1, \ldots, \theta_m}(\gamma) = \sum_{j=1}^m \left[ \theta_1(y) - \theta_1(x) \right] \hat{F}_{\theta_2, \ldots, \theta_j \cdot \theta_{j+1}, \ldots, \theta_m}(\gamma \setminus x)$$

(3.5)

$$+ \left( \exp(-\tau \psi(y)) - \exp(-\tau \psi(x)) \right) \hat{F}_{\theta_1, \ldots, \theta_m}(\gamma \setminus x).$$
A Markov process for a particle system

For $\theta \in \Theta_0$ and $a$ as in (3.2), we set

$$
(a \ast \theta)(x) = \int_{\mathbb{R}^d} a(x - y)\theta(y)dy = \int_{\mathbb{R}^d} \theta(x - y)a(y)dy. \quad (3.6)
$$

Then $a \ast \theta \in C_b(\mathbb{R}^d)$, where the continuity follows by the dominated convergence theorem and the latter equality in (3.6). Moreover, by (2.34) we have

$$
(a \ast \theta)(x) \leq \hat{c}_g\psi(x)\left[1 + \int_{\mathbb{R}^d} (|x - y| + |y|)^{d+1}a(x - y)\psi(y)dy\right]
$$

where we have used (3.2), (3.3) and the fact that $|y|^l\psi(y) \leq 1$ holding for all $y$ and $0 \leq l \leq d + 1$. Therefore,

$$
\theta_j^+(x) := (a \ast \theta_j)(x) + \theta_j(x) \leq c_a\hat{c}_\theta\psi(x). \quad (3.8)
$$

Since $\theta_j \in \Theta_0^+$, we then get by the latter that also $\theta_j^+ \in \Theta_0^+$, $j = 1, \ldots, m$. Here

$$
c_a := 2 + \sum_{l=0}^{d+1} \binom{d+1}{l}m_l^a. \quad (3.9)
$$

At the same time

$$
|\exp(-\tau\psi(y)) - \exp(-\tau\psi(x))| \leq \tau\psi(y)\psi(x)||x|^{d+1} - |y|^{d+1}|. \quad (3.10)
$$

Then proceeding as in (3.7) we get

$$
\int_{\mathbb{R}^d} a(x - y)\exp(-\tau\psi(y)) - \exp(-\tau\psi(x))dy \leq \tau c_a \psi(x). \quad (3.11)
$$

Thereafter, by (3.5), (3.6), (3.8), and (1.2) we obtain

$$
\left|L\hat{F}_\tau^{\theta_1, \ldots, \theta_m}(\gamma)\right| = \left|\sum_{\gamma \in \mathbb{G}} \int_{\mathbb{R}^d} a(x - y)\exp\left(-\sum_{z \in \gamma \setminus x} \phi(z - y)\right)\nabla y, x \hat{F}_\tau^{\theta_1, \ldots, \theta_m}(\gamma)dy\right| \quad (3.12)
$$

where, cf. (2.42),

$$
\hat{F}_\tau^{m}(\gamma) = \sum_{x_1 \in \gamma} \psi(x_1) \sum_{x_2 \in \gamma \setminus x_1} \psi(x_2) \cdots \sum_{x_m \in \gamma \setminus \{x_1, \ldots, x_{m-1}\}} \psi(x_m)\hat{F}_\tau^{0}(\gamma \setminus \{x_1, \ldots, x_m\}). \quad (3.13)
$$

Then the boundedness of $L\hat{F}_\tau^{\theta_1, \ldots, \theta_m}$ follows by Proposition 2.15.
A Markov process for a particle system

Now let us show that $L\tilde{F}^\theta_{\tau} \in \mathcal{B}_b(\Gamma_\ast)$ for all $\theta \in \Theta_\circ$ and $\tau > c_\theta$. Similarly as in (3.5) we get

$$\nabla^{\varphi,\varphi} \tilde{F}^\theta_{\tau}(\gamma) = \begin{bmatrix} e^{-\tau \psi(y)} - e^{-\tau \psi(x)} \\ \theta(y) e^{-\tau \psi(y)} - \theta(x) e^{-\tau \psi(x)} \end{bmatrix} \tilde{F}^\theta_{\tau}(\gamma \setminus x).$$

Then by (3.8), (3.9), (3.10) and (3.11) we arrive at

$$\left| L\tilde{F}^\theta_{\tau}(\gamma) \right| \leq e^r (1 + \tau) c_\theta \Psi(\gamma) \exp (-\tau_0 \Psi(\gamma)) \prod_{x \in \gamma} (1 + \theta(x)) e^{-(r - \tau_0) \psi(x)},$$

where $\tau_0 > 0$ is such that $\tau - \tau_0 > c_\theta$ which is possible for each $\tau > c_\theta$. Then the boundedness in question follows similarly as in Proposition 2.15. The next statement summarizes the properties of $\mathcal{D}(L)$.

**Proposition 3.2.** The set of functions introduced in Definition 3.1 has the following properties:

1. $\mathcal{D}(L) \subset C_b(\Gamma_\ast)$ and $L : \mathcal{D}(L) \to \mathcal{B}_b(\Gamma_\ast)$.
2. The set $\mathcal{B}_b(\Gamma_\ast)$ is the bp-closure of $\mathcal{D}(L)$.
3. $\mathcal{D}(L)$ is separating. That is, if $\mu_1, \mu_2 \in \mathcal{P}_{\exp}$ satisfy $\mu_1(F) = \mu_2(F)$ for all $F \in \mathcal{D}(L)$, then $\mu_1 = \mu_2$.
4. For each $F \in \tilde{\mathcal{F}}$, see (3.4), and $\mu \in \mathcal{P}_{\exp}$, the measure $F \mu / \mu(F)$ belongs to $\mathcal{P}_{\exp}$.

**Proof.** Claim (i) has been just proved. Claims (ii) and (iii) follow by Proposition 2.14 and the fact $\tilde{\mathcal{F}} \subset \mathcal{D}(L)$. It remains to check the validity of (2.8) for $\mu_F := F \mu / \mu(F)$. For positive $\theta \in C_{cs}(\mathbb{R}^d)$ and a bounded positive $F$, we have that

$$\chi_\mu(\theta^{\otimes n}) = \frac{1}{\mu(F)} \int_{\Gamma_\ast} F(\gamma) \left( \sum_{x_1 \in \gamma} \sum_{x_2 \in \gamma \setminus x_1} \cdots \int_{\gamma \setminus \{x_1, \ldots, x_{n-1}\}} \theta(x_1) \theta(x_2) \cdots \theta(x_n) \right) \mu(d\gamma) \leq \chi_\mu(\theta^{\otimes n}) \left( \sup_\gamma F(\gamma) / \mu(F) \right),$$

which completes the proof. \(\-boxed{}\)

### 3.2 Formulating the result

As mentioned in Introduction, following [13, Chapter 5] we are going to obtain the process by solving a restricted initial value martingale problem. Recall that $\mathcal{D}_{\exp}(\Gamma_\ast)$ stands for the space of all cadlag maps $[0, +\infty) \to \Gamma_\ast$ that are right-continuous and take values in $\Gamma_\ast$, and the evaluation maps $\varpi_t : t \mapsto \varpi_t(\gamma) = \gamma_t \in \Gamma_\ast$. In a similar way, one defines also the spaces $\mathcal{D}_{[s, +\infty)}(\Gamma_\ast)$, $s > 0$. For $s, t \geq 0, s < t$, by $\tilde{\mathcal{F}}_{s,t}^0$ we denote the $\sigma$-field of subsets of $\mathcal{D}_{[s,t]}(\Gamma_\ast)$ generated by the family $\{\varpi_u : u \in [s, t]\}$. Then we set

$$\tilde{\mathcal{F}}_{s,t} = \bigcap_{\varepsilon > 0} \tilde{\mathcal{F}}_{s,t+\varepsilon}, \quad \tilde{\mathcal{F}}_{s, +\infty} = \bigvee_{n \in \mathbb{N}} \tilde{\mathcal{F}}_{s, s+n}. \quad (3.14)$$

That is, $\tilde{\mathcal{F}}_{s, +\infty}$ is the smallest $\sigma$-field which contains all $\tilde{\mathcal{F}}_{s, s+n}$. Given $s \geq 0$ and $\mu \in \mathcal{P}_{\exp}$, in the definition below – which is an adaptation of the definition in [13, Section 5.1, pages 78, 79] – we deal with probability measures $P_{s, \mu}$ on $(\mathcal{D}_{[s, +\infty)}(\Gamma_\ast), \tilde{\mathcal{F}}_{s, +\infty})$.

**Definition 3.3.** A family of probability measures $\{P_{s, \mu} : s \geq 0, \mu \in \mathcal{P}_{\exp}\}$ is said to be a solution of the restricted initial value martingale problem for our model if, for all $s \geq 0$ and $\mu \in \mathcal{P}_{\exp}$, the following holds: (a) $P_{s, \mu} \circ \varpi_{s}^{-1} = \mu$; (b) $P_{s, \mu} \circ \varpi_{t}^{-1} \in \mathcal{P}_{\exp}$ for
A Markov process for a particle system

all \( t > s \); (c) for each \( F \in \mathcal{D}(L) \) (Definition 3.1), \( t_2 \geq t_1 \geq s \) and any bounded function \( G : \mathcal{D}_{[s, +\infty)}(\Gamma) \to \mathbb{R} \) which is \( \mathfrak{F}_{s, t_1} \)-measurable, the function

\[
H(\gamma) := \left[ F(\varpi_{t_2}(\gamma)) - F(\varpi_{t_1}(\gamma)) - \int_{t_1}^{t_2} (LF)(\varpi_u(\gamma))du \right] G(\gamma)
\]

is such that

\[
\int_{\mathcal{D}_{[s, +\infty)}} H(\gamma) P_{s, \mu}(d\gamma) = 0.
\]

The restricted initial value martingale problem is said to be well-posed if, for each \( s \geq 0 \) and \( \mu \in \mathcal{P}_{\exp} \), there exists a unique \( P_{s, \mu} \) satisfying all the conditions mentioned above.

Here by saying “for our model” along with the Kolmogorov operator \( L \) given in (1.2) we mean also its domain \( \mathcal{D}(L) \) (Definition 3.1) and the class \( \mathcal{P}_{\exp} \) defined by the property (2.8). Note that \( H \) defined in (3.15) is \( P_{s, \mu} \)-integrable, that follows by claim (i) of Proposition 3.2. Note also that the functions \( G \) in (3.15) can be taken in the form

\[
G(\gamma) = F_1(\varpi_{s_1}(\gamma)) \cdots F_m(\varpi_{s_m}(\gamma)),
\]

with all possible choices \( m \in \mathbb{N}, F_1, \ldots, F_m \in \hat{F} \) (see Proposition 2.14), and \( s \leq s_1 < s_2 < \cdots < s_m \leq t_1 \), see [15, eq. (3.4), page 174].

**Definition 3.4.** For a given \( s \geq 0 \), a map, \( [s, +\infty) \ni t \mapsto \mu_t \in \mathcal{P}(\Gamma_s) \), is said to be measurable if the maps \( [s, +\infty) \ni t \mapsto \mu_t(\mathcal{A}) \in \mathbb{R} \) are measurable for all \( \mathcal{A} \in \mathcal{B}(\Gamma_s) \). Such a map is said to be a solution of the Fokker-Planck equation for our model if, for each \( F \in \mathcal{D}(L) \) and any \( t_2 > t_1 \geq s \), the following holds

\[
\mu_{t_2}(F) = \mu_{t_1}(F) + \int_{t_1}^{t_2} \mu_u(LF)du.
\]

**Remark 3.5.** In view of the integral form of (3.18), its solutions are often called weak. We do not do this as the precise meaning of this notion is clear from the definition above. By taking \( G \equiv 1 \) in (3.15) one comes to the following conclusion. Let \( \{P_{s, \mu} : s \geq 0, \mu \in \mathcal{P}_{\exp}\} \) be a solution as in Definition 3.3. Then, for each \( s \) and \( \mu \in \mathcal{P}_{\exp} \), the map \( [s, +\infty) \ni t \mapsto P_{s, \mu} \circ \varpi_t^{-1} \) solves (3.18) for all \( t_2 > t_1 \geq s \).

Below, by \( \mathcal{D}_{[s, +\infty)}(\tilde{\Gamma}_s), s \geq 0 \), we mean the space of cadlag maps \( [s, +\infty) \ni t \mapsto \gamma_t \in \tilde{\Gamma}_s \), where the latter is the space of single configurations, see (2.30). Now we formulate our principal result.

**Theorem 3.6.** For the model defined in (1.2) satisfying (3.2) and (3.3), the following is true:

(a) **The restricted initial value martingale problem is well-posed in the sense of Definition 3.3.**

(b) **The stochastic process \( X \) related to the family**

\[
(\mathcal{D}_{[s, +\infty)}(\Gamma_s), \mathfrak{F}_{s, +\infty}, \{\mathfrak{F}_{s, t} : t \geq s\}, \{P_{s, \mu} : \mu \in \mathcal{P}_{\exp}\})_{s \geq 0}
\]

**is Markov.** This means that, for all \( t > s \) and \( \mathcal{B} \in \mathfrak{F}_{t, +\infty} \), the following holds

\[
\text{Prob}(X \in \mathcal{B}) = P_{s, \mu}(\mathcal{B} \mid \mathfrak{F}_{s, t}) = P_{s, \mu}(\mathcal{B} \mid \mathfrak{F}_{t}), \quad P_{s, \mu} - \text{almost surely}.
\]

Here \( \mathfrak{F}_t \) is the smallest \( \sigma \)-field of subsets of \( \mathcal{D}_{[s, +\infty)} \) that contains all \( \varpi_t^{-1}(\mathcal{A}) \), \( \mathcal{A} \in \mathcal{B}(\Gamma) \).
A Markov process for a particle system

(c) The aforementioned process has the property

\[ \operatorname{Prob}\left( X \in \mathcal{D}_{R^+}(\tilde{\Gamma}_*) \right) = 1. \]

The proof of claim (a) of this statement is the main concern of the rest of the paper. It will be done in the following two steps. First we prove that the restricted initial value martingale problem as in Definition 3.3 has at most one solution. Thereafter, we construct a solution by ‘superposing’ (cf. [32]) the collection of measures constructed in [4].

3.3 Strategy of the proof and some comments

Our approach is essentially based on the Fokker-Planck equation (1.3), (3.18) for which a solution, \( t \to \mu_t \in \mathcal{P}_{\exp}, \mu_0 \in \mathcal{P}_{\exp} \) was constructed in [4]. In Sect. 6, we introduce approximating models by modifying the jump kernel in such a way that allows one to solve the Fokker-Planck equation directly by constructing stochastic semigroups in a Banach space of signed measures, with the possibility to take Dirac measures \( \delta_\gamma, \gamma \in \tilde{\Gamma}_* \), as the initial conditions. This allows in turn for introducing finite-dimensional marginals of the presumed law of the processes corresponding to these approximating models by means of the transition functions obtained in that way. Then we prove that these marginals satisfy a Chentsov-like condition (see [15, Theorem 3.8.8, page 139]) – the same for all approximating models. Here we employ the complete metric of \( \Gamma_* \), see (2.24). This yields the existence of cadlag versions of the approximating processes and is used in Sect. 7 to prove that their distributions have accumulating points – possible distributions of the process in question. Then we prove that such accumulation points solve the martingale problem in the sense of Definition 3.3. To prove uniqueness we again use the Fokker-Planck equation and the construction made in [4]. At this stage – realized in Sect. 5 – we show that this equation has a unique solution, which implies that the mentioned accumulation points have coinciding one-dimensional marginals. A classical result (see [15, claim (a) of Theorem 4.4.2, page 184]) is that one would have uniqueness if the one-dimensional marginals were equal for all initial \( \mu \in \mathcal{P}(\tilde{\Gamma}_*) \). Since we have such an equality only for \( \mu \) from a subset of \( \mathcal{P}(\tilde{\Gamma}_*) \), we turn to the restricted version of the martingale problem [13, Chapter 5]. A crucial element of this version is Lemma 5.1 that states that a solution of the Fokker-Planck equation with \( \mu_0 \in \mathcal{P}_{\exp} \) is also in \( \mathcal{P}_{\exp} \), and its type satisfies \( \kappa_t \leq \kappa_T \) for \( t \leq T \), where \( \kappa_T \) depends on \( T \) and \( \kappa_0 \) only.

The proof of Lemma 5.1 is the most technical element of this part, based on a number of combinatorial results (see also Appendix). By means of Lemma 5.1 we then prove (Theorem 5.2) that (1.3) with \( \mu_0 \in \mathcal{P}_{\exp} \) has a unique solution coinciding with the map \( t \to \mu_t \) constructed in [4]. This finally yields the uniqueness of the solution.

4 The Evolution of States on \( \Gamma_* \)

As mentioned above, in the proof of Theorem 3.6 we essentially use the construction of the family of measures \( \{\mu_t\}_{t \geq 0} \subset \mathcal{P}_{\exp} \) performed in [4]. Notably, in this construction, there was used the space of single configurations \( \tilde{\Gamma}_* \), which for measures from \( \mathcal{P}_{\exp} \) makes no difference, see Remark 2.11. Thus, we begin by describing this family in a way adapted to the present context.

4.1 Spaces of functions on \( \Gamma_0 \)

By (2.19) it follows that each measurable \( F \) satisfying \( |F(\gamma)| \leq C \exp(\beta \Psi(\gamma)) \) for some positive \( \beta \) and \( C \) is \( \mu \)-absolutely integrable for each \( \mu \in \mathcal{P}_{\exp} \). This obviously relates to \( F = KG \) with \( G \in \mathcal{B}_{bs} \), see Remark 2.2. For \( a \) and \( \phi \) as in (3.2) and \( G \in \mathcal{B}_{bs} \), let us
A Markov process for a particle system

consider

\[
(\hat{L}G)(\eta) = \sum_{\xi \in \eta} \sum_{x \in \xi} \int_{\mathbb{R}^d} a(x - y)e(\tau_y; \xi)e(t_y; \eta \setminus \xi) [G(\xi \setminus x \cup y) - G(\xi)] \, dy, \tag{4.1}
\]

\[
\tau_y(x) := e^{-\phi(x-y)}, \quad t_y(x) := \tau_y(x) - 1, \quad x, y \in \mathbb{R}^d.
\]

In (4.1), the sums are finite and the integral is convergent in view of the integrability of the jump kernel \(a\). It turns out that

\[
LKG = K\hat{L}G,
\]

holding for all \(G \in B_{bs}\), see [16, Corollary 4.3 and eq. (4.7)]. By (2.5) this yields

\[
\mu(LKG) = \langle k_\mu, \hat{L}G \rangle, \tag{4.2}
\]

which by (2.9) points to the possibility to extend \(\hat{L}\) from \(B_{bs}\) to integrable functions. For a given \(\vartheta \in \mathbb{R}\), let \(G_\vartheta\) stand for the weighted \(L^1\)-space equipped with the norm

\[
|G|_\vartheta = \int_{\Gamma_0} |G(\eta)| \exp(\vartheta |\eta|) \lambda(d\eta)
\]

\[
= |G(\emptyset)| + \sum_{n=1}^{\infty} e^{\vartheta n} \int_{\mathbb{R}^n} |G^{(n)}(x_1, \ldots, x_n)| dx_1 \cdots dx_n.
\]

In fact, we have a descending scale \(\{G_\vartheta : \vartheta \in \mathbb{R}\}\) such that

\[
G_{\vartheta'} \hookrightarrow G_\vartheta, \quad \vartheta' > \vartheta,
\]

where by \(\hookrightarrow\) we mean continuous embedding. For a given \(\vartheta \in \mathbb{R}\) and \(G \in B_{bs}\), let us estimate \(|\hat{L}G|_\vartheta\). By means of [16, Lemma 2.3], see also [4, Lemma 3.1], by (3.2) we get

\[
|\hat{L}G|_\vartheta \leq \int_{\Gamma_0} e^{\vartheta |\eta|} \left( \sum_{\xi \in \eta} \sum_{x \in \xi} \int_{\mathbb{R}^d} a(x - y) \left( |G(\eta \setminus x \cup y)| + |G(\xi)| \right) + |G(\eta)| \right) \lambda(d\eta)
\]

\[
\leq 2 \exp(e^{\vartheta \phi}) \int_{\Gamma_0} e^{\vartheta |\eta|} |G(\eta)| \lambda(d\eta).
\]

To estimate the last line in the latter formula we use the inequality \(xe^{-\alpha x} \leq 1/e\alpha\), both \(x, \alpha\) positive, and the fact that \(B_{bs} \subset G_{\vartheta'}\) for each \(\vartheta' > \vartheta\). Thereafter, we obtain

\[
|\hat{L}G|_\vartheta \leq \frac{2}{e^{(\vartheta' - \vartheta)}} \exp(e^{\vartheta \phi}) |G|_{\vartheta'}.
\]

Below by means of this estimate we extend \(\hat{L}\) to operators acting in the scale \(\{G_\vartheta\}_{\vartheta \in \mathbb{R}}\), cf. (4.4).
Along with \( G_{\vartheta} \), we introduce the following Banach spaces. For symmetric \( k^{(n)} \in L^{\infty}((\mathbb{R}^d)^n), \ n \in \mathbb{N} \), let \( k \) be defined by \( k^{(n)} \) as in (2.2), that includes also some constant \( k(\varnothing) = k^{(0)} \). Such \( k \) constitute a real linear space and can be considered as essentially bounded functions \( k : \Gamma_0 \to \mathbb{R} \). Note that the correlation functions \( k_\mu \), cf. (2.9), are such functions. Then for \( \vartheta \in \mathbb{R} \), we define

\[
\|k\|_\vartheta = \sup_{n \geq 0} \\left( \|k^{(n)}\|_{L^{\infty}((\mathbb{R}^d)^n)} e^{-\vartheta n} \right) = \text{ess sup}_{\eta \in \Gamma_n} \left( |k(\eta)| \exp(-\vartheta |\eta|) \right).
\]

The linear space \( K_{\vartheta} \) equipped with this norm is the Banach space in question. Clearly, cf. (4.4),

\[
K_{\vartheta} \hookrightarrow K_{\vartheta'}, \quad \text{for} \ \vartheta < \vartheta'.
\]

Note that \( K_{\vartheta} \) is the topological dual to \( G_{\vartheta} \) as the value of \( k \) on \( G \) is given by the formula

\[
\langle k, G \rangle = \int_{\Gamma_n} k(\eta) G(\eta) \lambda(d\eta).
\]

Let us now define \( L^\Delta \) by the condition, cf. (4.2),

\[
\langle L^\Delta k_\mu, G \rangle = \langle k_\mu, L G \rangle.
\]

By (4.1) it is obtained in the following form, see [4, eqs. (2.21), (2.22)],

\[
(L^\Delta k)(\eta) = \sum_{y \not\in \eta} \int_{\mathbb{R}^d} a(x - y)e(\tau_y; \eta \setminus y \cup x) (W_y k)(\eta \setminus y \cup x) dx
\]

\[
- \sum_{x \in \eta} \int_{\mathbb{R}^d} a(x - y)e(\tau_y; \eta) (W_y k)(\eta) dy,
\]

where

\[
(W_y k)(\eta) = \int_{\Gamma_n} k(\eta \cup \xi)e(t_y; \xi) \lambda(d\xi).
\]

Proceeding similarly as in obtaining (4.5), for all \( \vartheta \in \mathbb{R} \) and \( \vartheta' > \vartheta \), we get

\[
\|L^\Delta k\|_{\vartheta'} \leq \frac{2}{e(\vartheta' - \vartheta)} \exp(e^{\vartheta}(\vartheta)) \|k\|_{\vartheta},
\]

where we have taken into account that \( \langle a \rangle = 1 \), see (3.2).

### 4.2 The evolution in spaces of functions on \( \Gamma_0 \)

By combining (4.2) with (4.7) we introduce the following versions of the Kolmogorov equation (1.1)

\[
\frac{d}{dt} G_t = \hat{L} G_t, \quad G_t|_{t=0} = G_0,
\]

\[
\frac{d}{dt} k_t = L^\Delta k_t, \quad k_t|_{t=0} = k_0,
\]

which we will solve in the scales \( \{G_{\vartheta} : \vartheta \in \mathbb{R}\} \) and \( \{K_{\vartheta} : \vartheta \in \mathbb{R}\} \), respectively, see (4.4) and (4.6).

Let us first consider (4.12). By (4.10) we see that \( L^\Delta \) maps each \( K_{\vartheta} \) in each \( K_{\vartheta'} \), cf. (4.6), and the corresponding map is linear and bounded. Likewise, one can define the linear maps \( (L^\Delta)^n : K_{\vartheta} \to K_{\vartheta'}, \ n \in \mathbb{N} \) the norm of which can be estimated by means of the inequality

\[
\|(L^\Delta)^n k\|_{\vartheta'} \leq \frac{n^n}{(e(T(\vartheta', \vartheta)))^n} \|k\|_{\vartheta},
\]
where, cf. [4, eq. (4.2)],

\[ T(\vartheta_2, \vartheta_1) = \frac{\vartheta_2 - \vartheta_1}{2} \exp \left( -\langle \phi \rangle e^{\vartheta_2} \right), \quad \vartheta_2 > \vartheta_1. \]  

(4.14)

It is known, see [4, eqs. (4.3), (4.4)] that

\[ \sup_{\vartheta' > \vartheta} T(\vartheta', \vartheta) = \frac{\delta(\vartheta)}{2} \exp \left( -\frac{1}{\delta(\vartheta)} \right) =: \tau(\vartheta), \]  

(4.15)

where \( \delta(\vartheta) \) is a unique solution of \( \delta e^\delta = e^{-\vartheta} / \langle \phi \rangle \). The supremum in (4.15) is attained at \( \vartheta' = \vartheta + \delta(\vartheta) \). Then the expression in (4.8), (4.9) can be used to define: (a) bounded linear operators \( (L^\Delta)_{\vartheta', \vartheta} : \mathcal{K}_\vartheta \to \mathcal{K}_{\vartheta'}, n \in \mathbb{N} \) the norm of which can be estimated by means of (4.13); (b) unbounded linear operators \( L^\Delta_{\vartheta'} \) with domains, cf. [4, eq. 3.19]],

\[ \text{Dom} L^\Delta_{\vartheta'} = \{ k \in \mathcal{K}_{\vartheta'} : L^\Delta k \in \mathcal{K}_{\vartheta} \}. \]

Now we turn to (4.11). In a similar way, by means of (4.5) one defines: (a) bounded linear operators \( (\hat{L})_{\vartheta', \vartheta}^n : \mathcal{G}_{\vartheta'} \to \mathcal{G}_{\vartheta}, n \in \mathbb{N} \), the norm of which satisfies

\[ \| (\hat{L})_{\vartheta', \vartheta}^n \| = \| (L^\Delta)_{\vartheta', \vartheta}^n \|, \quad n \in \mathbb{N}; \]  

(4.16)

(b) unbounded operators \( \hat{L}_{\vartheta} \) with domains

\[ \text{Dom} \hat{L}_{\vartheta} = \{ G \in \mathcal{G}_{\vartheta} : \hat{L}G \in \mathcal{G}_{\vartheta} \}. \]

It can be shown, see [4, Lemma 3.1], that, for each \( \vartheta \in \mathbb{R} \) and \( \vartheta' > \vartheta \), the following is true

\[ \mathcal{K}_{\vartheta} \subset \text{Dom} L^\Delta_{\vartheta'}, \quad \mathcal{G}_{\vartheta'} \subset \text{Dom} \hat{L}_{\vartheta}, \]

by which one readily obtains that, for all \( \vartheta, \vartheta', \vartheta'' > \vartheta \), the following holds

\[ \forall k \in \mathcal{K}_{\vartheta} \quad L^\Delta_{\vartheta', \vartheta} k = L^\Delta_{\vartheta'' \vartheta} k. \]  

(4.17)

Furthermore, up to the embedding (4.6) we have that

\[ L^\Delta_{\vartheta', \vartheta} k = L^\Delta_{\vartheta'' \vartheta} k, \]

holding for all \( \vartheta'' \in (\vartheta, \vartheta') \). By (4.13) the series

\[ Q_{\vartheta', \vartheta}(t) = 1 + \sum_{n=1}^{\infty} \frac{t^n}{n!} (L^\Delta)_{\vartheta', \vartheta}^n \]  

(4.18)

converges in the operator norm topology – uniformly on compact subsets of \([0, T(\vartheta' \vartheta))\) – to a bounded linear operator

\[ Q_{\vartheta', \vartheta}(t) : \mathcal{K}_{\vartheta} \to \text{Dom} L^\Delta_{\vartheta'}, \subset \mathcal{K}_{\vartheta'}, \]

the norm of which satisfies

\[ \| Q_{\vartheta', \vartheta}(t) \| \leq \frac{T(\vartheta', \vartheta)}{T(\vartheta', \vartheta) - t}. \]  

(4.19)

Moreover, the map \([0, T(\vartheta' \vartheta)) \ni t \mapsto Q_{\vartheta', \vartheta}(t)\) is differentiable and the following holds

\[ \frac{d}{dt} Q_{\vartheta', \vartheta}(t) = L^\Delta_{\vartheta', \vartheta} Q_{\vartheta', \vartheta}(t) = L^\Delta_{\vartheta'', \vartheta} Q_{\vartheta', \vartheta}(t) = Q_{\vartheta', \vartheta''}(t) L^\Delta_{\vartheta'', \vartheta}, \]  

(4.20)

with an arbitrary \( \vartheta'' \in (\vartheta, \vartheta') \) provided \( t \) satisfies \( t < T(\vartheta'', \vartheta) \) and \( t < T(\vartheta', \vartheta'') \) in the latter two terms, respectively, cf. (4.19). By (4.20) one readily obtains that the Cauchy
A Markov process for a particle system

problem in (4.12) with \( k_0 \in \mathcal{K}_\vartheta \) has a unique classical solution in \( \mathcal{K}_\vartheta \), on the time interval \([0, T(\vartheta', \vartheta))\), cf. [4, Lemma 4.1]. It is

\[
k_t = Q_{\vartheta; \vartheta}(t)k_0. \tag{4.21}
\]

In a similar way, one shows that the Cauchy problem in (4.11) has a unique classical solution in \( \mathcal{G}_\vartheta \), on the time interval \([0, T(\vartheta', \vartheta))\), given by the formula

\[
G_t = H_{\vartheta; \vartheta'}(t)G_0 = \left(1 + \sum_{n=1}^{\infty} \frac{t^n}{n!} (\hat{L})_{\vartheta; \vartheta'}^n \right) G_0, \quad G_0 \in \mathcal{G}_{\vartheta'}. \tag{4.22}
\]

By construction these solutions of (4.12) and (4.11) satisfy

\[
\langle \langle k_t, G_0 \rangle \rangle = \langle \langle k_0, G_t \rangle \rangle, \quad t < T(\vartheta', \vartheta). \tag{4.23}
\]

4.3 The evolution of states

A priori, the solution given in (4.21) need not be the correlation function for any measure. Moreover, it may not even be positive, cf. (2.9). To check whether a given \( k : \Gamma_0 \to \mathbb{R}^+ \) is the correlation function of a certain \( \mu \in \mathcal{P}_{\exp} \) we introduce the following set

\[
B_{bs}^* = \{ G \in B_{bs} : \sum_{\xi \subset \eta} G(\xi) \geq 0, \text{ for all } \eta \in \Gamma_0 \}. \tag{4.24}
\]

Note that some of its members can take also negative values. By [21, Theorems 6.1 and 6.2 and Remark 6.3] one proves the following statement.

**Proposition 4.1.** Let a measurable function, \( k : \Gamma_0 \to \mathbb{R} \), have the following properties:

(a) \( \langle \langle k, G \rangle \rangle \geq 0 \) for all \( G \in B_{bs}^* \), see (4.24);
(b) \( k(\emptyset) = 1 \);
(c) \( k(\eta) \leq \kappa|\eta| \) for some \( \kappa > 0 \), cf. (2.9).

Then \( k \) is the correlation function for a unique \( \mu \in \mathcal{P}_{\exp} \).

Recall that the least \( \kappa \) as in item (c) above is the type of \( \mu \) of which \( k \) is then the correlation function. Set

\[
\mathcal{P}^\vartheta_{\exp} = \{ \mu \in \mathcal{P}_{\exp} : \mu \text{ is of type } \leq e^\vartheta \}. \tag{4.25}
\]

Let \( \mathcal{K}^* \) be the set of all \( k : \Gamma_0 \to \mathbb{R} \) that possess the properties listed in Proposition 4.1. In [4, Theorem 3.3], it was shown that \( k_t \) as in (4.21) belongs to \( \mathcal{K}^* \) whenever \( k_0 \) is the correlation function of a certain \( \mu \in \mathcal{P}_{\exp} \). In the context of the present study, the relevant results of [4] can be formulated as follows.

**Proposition 4.2.** Given \( \vartheta_0 \in \mathbb{R} \), let \( \mu \) be an arbitrary element of \( \mathcal{P}^\vartheta_{\exp} \). For this \( \vartheta_0 \), set \( \vartheta_t = \vartheta_0 + t, \ t \geq 0 \). Then there exists a unique map, \( [0, +\infty) \ni t \mapsto k_t \in \mathcal{K}^* \), such that \( k_0 = k_\mu \) and the following holds:

(a) for each \( t > 0 \),

\[
0 \leq k_t(\eta) \leq e^{\vartheta_t|\eta|}, \quad \eta \in \Gamma_0,
\]

by which \( k_t \in \mathcal{K}_{\vartheta_t} \).

(b) For each \( T > 0 \) and \( t \in [0, T) \), the map \( t \mapsto k_t \in \mathcal{K}_{\vartheta_t} \subset \text{Dom}L^A_{\vartheta_t} \) is continuous on \([0, T)\) and continuously differentiable on \([0, T)\) in \( \mathcal{K}_{\vartheta_T} \) and the following holds:

\[
\frac{d}{dt}k_t = L^A_{\vartheta_T}k_t. \tag{4.26}
\]
A Markov process for a particle system

5 The Uniqueness

In this section, we prove that the restricted initial value martingale problem has at most one solution. To this end we use the properties of $\mathcal{D}(L)$ stated in Proposition 3.2. In view of Remark 3.5, see also Lemma 5.4 below, the proof of the uniqueness in question amounts to proving that, for each $\mu \in \mathcal{P}_{\text{exp}}$, the Fokker-Planck equation (3.18) has at most one solution $\mu_t \in \mathcal{P}_{\text{exp}}$ satisfying $\mu_0 = \mu$. The main tool for this is based on controlling the type of $\mu_t$ by a method based on the use of the concrete form of the elements of $\mathcal{D}(L)$, see Definition 3.1.

5.1 Solving the Fokker-Planck equation

We begin by pointing out that in Definition 3.4 we do not assume that $\mu_t \in \mathcal{P}_{\text{exp}}$ for $t > 0$. Recall that $(KG)(\gamma) = \sum_{\xi \in \gamma} G(\xi)$, see (2.4).

**Lemma 5.1.** Let $[0, +\infty) \ni t \mapsto \mu_t \in \mathcal{P}(\Gamma_*)$ be a solution of (3.18) with all $F$ belonging to the linear span of $\mathcal{F}$ and a given $\mu_0 \in \mathcal{P}_{\text{exp}}$. Then, for each $T > 0$, there exists $\vartheta_T \in \mathbb{R}$ such that, for all $t \in [0, T]$, $\mu_t \in \mathcal{P}_{\text{exp}}^{\vartheta_t}$ with some $\vartheta_t < \vartheta_T$.

Note that here we assume that only the initial state $\mu_0$ belongs to $\mathcal{P}_{\text{exp}}$. Also, we assume that $\mu_t$ solves (3.18) with $F$ belonging only to a subset of $\mathcal{D}(L)$. It turns out that this is enough to solve it for all $\mathcal{D}(L)$, and even more. Set

$$\mathcal{F} = \left\{ F \in B_b(\Gamma_*): F = KG, \ G \in \bigcap_{\vartheta \in \mathbb{R}} \mathcal{G}_\vartheta \right\},$$

(5.1)

where $K$ is defined in (2.4) and $G$ is supposed to be such that $|G|_0$ is finite for all $\vartheta$, see (4.3). Let us show that $\mathcal{D}(L) \subset \mathcal{F}$. Since $K$ is linear, this will follow from the fact that

$$\mathcal{F} \cup \hat{\mathcal{F}} \subset \mathcal{F}.$$  

(5.2)

By (2.36) we have

$$\hat{F}_T^\vartheta(\gamma) = \prod_{x \in \gamma} (1 + \theta(x))e^{-\tau \psi(x)} = \sum_{\eta \subset \gamma} e(\theta_T; \eta) =: (KG_T^\vartheta)(\gamma),$$

(5.3)

$$\theta_T(x) := \theta(x)e^{-\tau \psi(x)} + \psi_T(x), \quad \psi_T(x) = -1 + e^{-\tau \psi(x)}. \quad \text{Clearly, } \theta_T \in L^1(\mathbb{R}^d) \text{ for each } \tau \geq 0 \text{ and } \theta \in \Theta_\psi, \text{ cf. Definition 3.1. Then } \hat{G}_T^\vartheta = e(\theta_T; \cdot) \in \mathcal{G}_\vartheta \text{ for any } \vartheta \in \mathbb{R}, \text{ which yields } \hat{F}_T \subset \mathcal{F}.

In the case of $F$ given in (2.42), (2.43), we write

$$\hat{F}_T^{\theta_1, \ldots, \theta_m}(\gamma) = \sum_{\xi \subset \gamma} g_m(\xi) \prod_{x \in \gamma \setminus \xi} (1 + \psi_T(x))$$

(5.4)

$$= \sum_{\eta \subset \gamma} \left( \sum_{\xi \subset \eta} g_m(\xi) e(\psi_T; \eta \setminus \xi) \right) =: \sum_{\eta \subset \gamma} \hat{G}_T^{\theta_1, \ldots, \theta_m}(\eta),$$

where $\psi_T(x)$ is as in (5.3) and

$$g_m(\xi) = \begin{cases} \sum_{\sigma \in S_m} \theta_1(x_{\sigma(1)}) \cdots \theta_m(x_{\sigma(m)}), & \text{if } \xi = \{x_1, \ldots, x_m\}; \\ 0 & \text{otherwise.} \end{cases}$$

Let us estimate $\hat{G}_T^{\theta_1, \ldots, \theta_m}$ with $\theta_1, \ldots, \theta_m \in \Theta_\psi^+$. For $\tau \in (0, 1]$, we have $|\psi_T(x)| \leq \psi(x)$, and hence

$$\left| \hat{G}_T^{\theta_1, \ldots, \theta_m}(\eta) \right| \leq \sum_{\xi \subset \eta} g_m(\xi) e(\psi; \eta \setminus \xi).$$

(5.5)
A Markov process for a particle system

At the same time, for each \( \eta \in \Gamma_0 \), it follows that

\[
\hat{G}_{\tau}^{\theta_0, \ldots, \theta_m}(\eta) \to g_m(\eta), \quad \tau \to 0^+.
\]  

(5.6)

By (5.5) let us show that \( \hat{G}_{\tau}^{\theta_1, \ldots, \theta_m} \) belongs to \( G_\vartheta \), \( \vartheta \in \mathbb{R} \). Indeed, by (4.3) we have

\[
|\hat{G}_{\tau}^{\theta_1, \ldots, \theta_m}|_\vartheta = \int_{\Gamma_0} |\hat{G}_{\tau}^{\theta_1, \ldots, \theta_m}(\eta)| e^{\vartheta(\eta)} \lambda(d\eta)
\]

where \( \langle \psi \rangle \), \( \langle \theta_i \rangle \), \( i = 1, \ldots, m \) are the corresponding \( L^1 \)-norms, cf. (2.16). This completes the proof of (5.2).

Lemma 5.1 is proved below. Now assuming that its claim holds true, we prove the next statement – one of the two basic tools of proving Theorem 3.6.

**Theorem 5.2.** For each \( \mu_0 \in \mathcal{P}_{\exp} \), the solution of the Fokker-Planck equation in the sense of Definition 3.4 exists and is unique.

**Proof.** We begin by showing that (3.18) has a solution. Take \( G = \hat{G}_{\tau}^{\theta_1, \ldots, \theta_m} \) and let \( k_\vartheta \) be as in Proposition 4.2 with \( k_0 \) being the correlation function of the initial state \( \mu_0 \). Since \( k_\vartheta \) is in \( K^* \), by Proposition 4.1 it determines a unique \( \mu_t \in \mathcal{P}_{\exp}^\theta \), see (4.25), for which

\[
\mu_t(F) = \mu_t(KG) = \langle \langle k_\vartheta(G) \rangle \rangle, \quad t \geq 0,
\]  

(5.8)

holding for all \( F \in \hat{F} \). By claim (a) of Proposition 4.2 and (5.1), (5.2) the integral in the right-hand side of (5.8) is absolutely convergent for each \( t \geq 0 \). Moreover, by claim (b) of Proposition 4.2 we have that

\[
k_{t_2} - k_{t_1} = \int_{t_1}^{t_2} L_{\hat{\vartheta}_u} k_u du
\]

holding for all \( t_2 > t_1 \geq 0 \) and \( T > t_2 \). We multiply both parts of the latter equality by an arbitrary \( G \in \cap_{\vartheta \in \mathbb{R}} G_\vartheta \) – also corresponding to \( F \in \hat{F} \) – and then integrate with respect to \( \lambda \). By claim (b) of Proposition 4.2 this integration and that over \( [t_1, t_2] \) can be interchanged, that implies

\[
\mu_{t_2}(F) - \mu_{t_1}(F) = \int_{t_1}^{t_2} \langle [L_{\hat{\vartheta}_u}^\lambda k_u, G] \rangle du = \int_{t_1}^{t_2} \langle [k_u, \hat{L} G] \rangle du = \int_{t_1}^{t_2} \mu_u(LF) du,
\]

(5.9)

where we have used (4.2), (4.7) and the fact that \( G \in \cap_{\vartheta \in \mathbb{R}} G_\vartheta \). This yields (3.18). By (5.2) we then get that \( \mu_t \) corresponding to \( k_\vartheta \) is a solution.

Assume now that there exists another solution, say \( \{\tilde{\mu}_t\}_{t \geq 0} \subset \mathcal{P}(\Gamma^*) \), such that \( \tilde{\mu}_0 = \mu_0 \). By Lemma 5.1 we have that \( \tilde{\mu}_t \in \mathcal{P}_{\exp}^{\tilde{\vartheta}_t} \) and \( \tilde{\vartheta}_t \in (\vartheta_0, \vartheta_T) \) for some \( \vartheta_T \) and all \( t \leq T \). This means that the corresponding correlation functions, \( k_\vartheta \), \( t \leq T \) belong to \( K_{\tilde{\vartheta}_t} \). Then the vector \( q_u = L_{\hat{\vartheta}_u}^\lambda \tilde{k}_u = L_{\hat{\vartheta}_u}^\lambda \tilde{k}_u \), see (4.17), lies in \( K_{\tilde{\vartheta}_u} \) and hence in \( K_{\tilde{\vartheta}_u + \varepsilon} \) for each \( \varepsilon > 0 \), see (4.6). Then, for a fixed \( \varepsilon \), by (4.13) and (2.9) we have

\[
\| q_u \|_{\tilde{\vartheta}_u + \varepsilon} \leq C(T, \varepsilon) e^{\tilde{\vartheta}_T}, \quad u \in [0, t],
\]  

(5.10)
with \( C(T, \varepsilon) = 1/\varepsilon T(\tilde{\vartheta}_T + \varepsilon, \tilde{\vartheta}_T) \), see (4.14). Let us prove that the following holds

\[
\forall G \in \bigcap_{\vartheta \in \mathbb{R}} \mathcal{G}_{\vartheta} \quad \langle \langle \tilde{k}_t - k_0, G \rangle \rangle = \int_0^t \langle \langle q_u, G \rangle \rangle du. \tag{5.11}
\]

A priori, the equality in (5.11) holds for only for \( G \) corresponding to \( F \in \mathcal{D}(L) \), that includes \( G = \widehat{G}_{\vartheta_1, \ldots, \vartheta_m} \), see (5.4). For \( \tau \in (0, 1] \), by (5.7) and (5.10) we then have

\[
\left| \langle \langle \tilde{k}_t - k_0, \widehat{G}_{\vartheta_1, \ldots, \vartheta_m} \rangle \rangle \right| \leq \langle \langle \tilde{k}_t - k_0 \rangle \| \delta_{\tau} \delta_{\vartheta_1} \cdots \delta_{\vartheta_m} (\tilde{\vartheta}_T + \varepsilon) \rangle \leq 2e^{\tilde{\vartheta}_T} \delta_{\vartheta_1} \cdots \delta_{\vartheta_m} (\tilde{\vartheta}_T + \varepsilon),
\]

\[
\int_0^t \langle \langle q_u, \widehat{G}_{\vartheta_1, \ldots, \vartheta_m} \rangle \rangle du \leq \int_0^t \langle \langle q_u \rangle \| \delta_{\vartheta_1} \cdots \delta_{\vartheta_m} (\tilde{\vartheta}_T + \varepsilon) \rangle \rangle du \leq tC(T, \varepsilon) e^{\tilde{\vartheta}_T} \delta_{\vartheta_1} \cdots \delta_{\vartheta_m} (\tilde{\vartheta}_T + \varepsilon).
\]

Now we write (5.11) for \( G = \widehat{G}_{\vartheta_1, \ldots, \vartheta_m} \) and pass to the limit \( \tau \to 0^+ \). By the dominated convergence theorem and (5.6) we then obtain

\[
\int_{(\mathbb{R}^d)^m} \left[ \tilde{k}_t^{(m)}(x_1, \ldots, x_m) - k_0^{(m)}(x_1, \ldots, x_m) \right] \theta_1(x_1) \cdots \theta_m(x_m) dx_1 \cdots dx_m \tag{5.12}
\]

that holds for all \( m \in \mathbb{N} \) and \( \theta_1, \ldots, \theta_m \in \Theta_+^\varepsilon \), see (2.32). For a fixed \( m \in \mathbb{N} \), the set of functions \( (x_1, \ldots, x_m) \mapsto \theta_1(x_1) \cdots \theta_m(x_m) \) with \( \theta_1, \ldots, \theta_m \in \Theta_+^\varepsilon \) is closed with respect to the pointwise multiplication and separates points of \( (\mathbb{R}^d)^m \). Such functions vanish at infinity and are everywhere positive. Then by the corresponding version of the Stone-Weierstrass theorem [10] the linear span of this set is dense (in the supremum norm) in the algebra \( C_0((\mathbb{R}^d)^m) \) of all continuous functions that vanish at infinity. At the same time, \( C_0((\mathbb{R}^d)^m) \cap L^1((\mathbb{R}^d)^m) \) is dense in \( L^1((\mathbb{R}^d)^m) \). For its subset \( C_{\alpha}((\mathbb{R}^d)^m) \) has this property. This allows us to extend the equality in (5.12) to the following

\[
\int_{(\mathbb{R}^d)^m} \left[ \tilde{k}_t^{(m)}(x_1, \ldots, x_m) - k_0^{(m)}(x_1, \ldots, x_m) \right] G^{(m)}(x_1, \ldots, x_m) dx_1 \cdots dx_m = \int_0^t \left( \int_{(\mathbb{R}^d)^m} q_u^{(m)}(x_1, \ldots, x_m) G^{(m)}(x_1, \ldots, x_m) dx_1 \cdots dx_m \right) du,
\]

holding for all \( G^{(m)} \in L^1((\mathbb{R}^d)^m) \). Then the passage from this equality to that in (5.11) follows by the fact that \( G \) belongs to each \( \mathcal{G}_{\vartheta} \), \( \vartheta \in \mathbb{R} \).

By (4.7) the equality in (5.11) yields

\[
\langle \langle \tilde{k}_t, G \rangle \rangle = \langle \langle k_0, G \rangle \rangle + \int_0^t \langle \langle \tilde{k}_u, \tilde{L}_{\tilde{\vartheta}, \vartheta} G \rangle \rangle du, \tag{5.13}
\]

in which \( \tilde{L}_{\tilde{\vartheta}, \vartheta} G =: G_1 \in \bigcap_{\vartheta \in \mathbb{R}} \mathcal{G}_{\vartheta} \). In view of (5.11), we can repeat (5.13) with \( G_1 \) instead of \( G \), and repeat this procedure again by employing the same arguments. After repeating \( n \) times we arrive at

\[
\langle \langle \tilde{k}_t, G \rangle \rangle = \langle \langle k_0, G \rangle \rangle + t \langle \langle k_0, \tilde{L}_{\tilde{\vartheta}, \vartheta} G \rangle \rangle + \frac{t^2}{2} \langle \langle k_0, (\tilde{L}_{\tilde{\vartheta}, \vartheta})^2 G \rangle \rangle + \cdots + \frac{t^{n-1}}{(n-1)!} \langle \langle k_0, (\tilde{L}_{\tilde{\vartheta}, \vartheta})^{n-1} G \rangle \rangle + \int_0^t \int_0^{t_1} \cdots \int_0^{t_n} \langle \langle \tilde{k}_{u_1}, (\tilde{L}_{\tilde{\vartheta}, \vartheta})^{n} G \rangle \rangle dt_1 \cdots dt_n.
\]
A Markov process for a particle system

Assume now that \( \tilde{\vartheta}_T > \vartheta_0 + T \), see Proposition 4.2, that is clearly possible by (4.6). Then we write down the same formula – in the same spaces – for \( k_t \) considered in (5.8), i.e., described in Proposition 4.2. This yields

\[
\langle k_t - k_t, G \rangle = \int_0^t \int_0^t \cdots \int_0^{t_{n-1}} \langle k_{t_n} - k_{t_n}, (L_{\vartheta_{t_i}})G \rangle dt_1 \cdots dt_n.
\]

Now we take \( \vartheta = \vartheta_1 + \delta(\vartheta_1) \), see (4.15). Then by (4.13), (4.16) and (4.15) we have from the latter

\[
\left| \langle k_t - k_t, G \rangle \right| \leq \frac{n^n}{n!} \left( \frac{t}{\tau(\vartheta_i)} \right)^n |G|_{\vartheta} \sup_{u \in [0, t]} \left( \|k_u\|_{\vartheta_i} + \|k_u\|_{\vartheta_i} \right). \tag{5.14}
\]

Note that here \( \tau(\vartheta_i) \geq \tau(\tilde{\vartheta}_T) \). Then for \( t < \tau(\tilde{\vartheta}_T) \), the right-hand side of (5.14) can be made as small as one wants by taking big enough \( n \). Since \( G \in \mathcal{G}_0 \) is arbitrary, this yields \( \hat{k}_t = k_t \) for all such \( t \). The latter implies \( \mu_t = \mu_t \), see Proposition 4.2. The continuation to bigger values of \( t \) is made by repeating the same procedure. The proof that these continuations cover the whole \( \mathbb{R}_+ \) can be done similarly as in the proof of Theorem 3.3 [4].

\[ \square \]

**Corollary 5.3.** Let \( t \mapsto \mu_t \) satisfy the assumptions of Lemma 5.1. Then it solves (3.18) with all \( F = KG \) with \( G \in \cap_{\vartheta \in \mathbb{R}} \mathcal{G}_0 \), also for unbounded ones.

**Proof.** By Lemma 5.1 a solution \( \mu_t \) is in \( \mathcal{P}^\vartheta_{\exp} \) for \( t \leq T \). Let \( k_t \) be its correlation function, which satisfies the equality in (5.11) with \( G = \tilde{\mathcal{G}}^\vartheta_1 \cdots \vartheta_m \). As we have shown in the proof of Theorem 5.2 it satisfies this equality for all \( G \) such that \( F = KG \) with \( G \in \cap_{\vartheta \in \mathbb{R}} \mathcal{G}_0 \), see (5.11). This yields the proof. \[ \square \]

### 5.2 Further properties of the solutions

In this subsection, we prepare proving Lemma 5.1. Our ultimate goal here is to estimate the integrals of the solutions of (3.18) taken with the functions

\[
F^\vartheta_m(\gamma) = \sum_{x_1 \in \gamma} \theta(x_1) \sum_{x_2 \in \gamma \setminus x_1} \theta(x_2) \cdots \sum_{x_m \in \gamma \setminus \{x_1, \ldots, x_{m-1}\}} \theta(x_m), \quad \theta \in \Theta_0^+, \tag{5.15}
\]

which can be obtained from the functions defined in (2.42) by setting \( \theta_1 = \cdots = \theta_m = \theta \) and \( \tau = 0 \). Note that \( F^\vartheta_m \) is unbounded, but integrable for each \( \mu \in \mathcal{P}^\vartheta_{\exp} \), as it follows from the formula, see (2.7),

\[
\mu(F^\vartheta_m) = \int_{\mathbb{R}^m} k^{(m)}_{\mu}(x_1, \ldots, x_m)\theta(x_1) \cdots \theta(x_m)dx_1 \cdots dx_m. \tag{5.16}
\]

Then by estimating \( \mu_t(F^\vartheta_m) \) we will prove the mentioned lemma.

To simplify notations by \( \Phi^m_\tau \) we denote a particular case of the function defined in (2.42), corresponding to the choice \( \theta_1 = \cdots = \theta_m = \theta \in \Theta^+_0 \) with \( \tau = 1 \), see (2.34). Namely, for \( \theta \in \Theta_0^+ \), we set, cf. also (5.15),

\[
\Phi^m_\tau(\gamma) = \sum_{x_1 \in \gamma} \theta(x_1) \sum_{x_2 \in \gamma \setminus x_1} \theta(x_2) \cdots \sum_{x_m \in \gamma \setminus \{x_1, \ldots, x_{m-1}\}} \theta(x_m)\tilde{F}^\vartheta_\tau(g \setminus \{x_1, \ldots, x_m\}), \tag{5.17}
\]

and consider such functions with \( \tau \in (0, 1] \). Note that the function defined in (3.13) is a particular case of \( \Phi^m_\tau(\gamma) \) corresponding to the choice \( \theta = \psi \). Then by (3.12) we obtain

\[
|L\Phi^m_\tau(\gamma)| \leq m\Phi^m_\tau(\gamma) + \tau e_\vartheta F^{m+1}_\tau(\gamma) =: \Phi^m_{\tau,1}(\gamma). \tag{5.18}
\]
Here $\theta^1 = a \ast \vartheta + \theta$, see (3.8), and
\[ \phi_{\tau}^{m,\vartheta} = \tilde{F}_{\tau}^{\vartheta_1,\vartheta_2,\ldots,\vartheta_m}, \quad \theta_2 = \cdots = \theta_m = \theta. \] (5.19)

Note that $\Phi_{\tau,1}^m$ is a linear combination of the elements of $\tilde{F}$, see (3.4). Hence, any solution of (3.18) should satisfy it also with this function. Let us then estimate $L\Phi_{\tau,1}^m$. Proceeding as in (3.5) we obtain
\[
\nabla^y \cdot \Phi_{\tau,1}^m(\gamma) = m[\theta^1(y) - \theta^1(x)]\Phi_{\tau,1}^{m-1}(\gamma \setminus x) + m(m-1)[\theta(y) - \theta(x)]\Phi_{\tau,1}^{m-1,\vartheta}(\gamma \setminus x) \\
+ m \left[ e^{-\tau \psi(y)} - e^{-\tau \psi(x)} \right] \Phi_{\tau,1}^{m,\vartheta} (\gamma \setminus x) + (m+1)\tau c_a[\psi(y) - \psi(x)] \tilde{F}_{\tau}^m(\gamma \setminus x) \\
+ \tau c_a \left[ e^{-\tau \psi(y)} - e^{-\tau \psi(x)} \right] \tilde{F}_{\tau}^{m+1}(\gamma \setminus x).
\]

Now to estimate $L\Phi_{\tau,1}^m$ we perform the same calculations as in passing to the second line in the right-hand side of (3.12), see (3.10), (3.11). In addition, the third term in the right-hand side of (3.18) should satisfy it also with this function. Let us then estimate $\Phi_{\tau}^m$. Note that
\[ \Psi \text{ and } \Phi \] are nonnegative integers, for $\gamma \in C_{\tau}^m$, we denote the set of all sequences $c = \{c_k\}_{k \in \mathbb{N}_0} \subset \mathbb{N}_0$ such that the following holds:
\[ c_0 + c_1 + \cdots + c_k + \cdots = m, \quad c_1 + 2c_2 + \cdots + kc_k + \cdots = n. \] (5.22)

Since all $c_k$ are nonnegative integers, for $c \in C_{\tau}^m$, by (5.22) we have that $c_{n+j} = 0$ for all $j \geq 1, c_n \leq 1$, and $c_j = 0$ for all $j = 1, 2, \ldots, n-1$ whenever $c_n = 1$. For example, $C_{\tau,0}$ and $C_{\tau,1}$ are singletons, consisting of $c = (m,0,0,\ldots)$ and $c = (m-1,1,0,\ldots)$, respectively. $C_{\tau,2}$ consists of $c = (m-1,0,1,0,\ldots)$ and $c = (m-2,2,0,\ldots)$. For $c \in C_{\tau,n}$, $\tau \in (0,1)$ and $\gamma \in \Gamma_+$, we set
\[ V_\tau(c;\gamma) = \tilde{F}_{\tau}^{\vartheta_1,\vartheta_2,\ldots,\vartheta_m}(\gamma), \] (5.23)
where $c_0$ members of the family $\{\vartheta_1,\ldots,\vartheta_m\}$ are equal to $\theta^0 = \theta$, $c_1$ of them are equal to $\theta^1$, etc. In particular, $\Phi_{\tau}^{m,\theta^2}$ and $\Phi_{\tau}^{m,\vartheta^3,\vartheta^4}$ can be written as in (5.23) with $c = (m-1,0,1,0,\ldots)$ and $c = (m-2,2,0,\ldots)$, respectively. In Appendix below, we prove the following estimates
\[ \forall \gamma \in \Gamma_+ \quad |L\Phi_{\tau,n-1}^m(\gamma)| \leq \Phi_{\tau,n}^m(\gamma), \quad n \in \mathbb{N}, \] (5.24)
A Markov process for a particle system

holding with $\Phi_{\tau,n}^m$ given by the following formula

$$\Phi_{\tau,n}^m(\gamma) = \sum_{c \in C_{m,n}} C_{m,n}(c)V_{\tau}(c; \gamma) + c_n^m \sum_{k=1}^{m} \tau^k w_k(m,n) \hat{F}_{\tau}^{m+k}(\gamma), \quad (5.25)$$

$$C_{m,n}(c) = \frac{m!n!}{c_0!c_1! \cdots c_k! \cdots (0!)^m(1)! \cdots (k!)^k \cdots}.$$  

We also prove that

$$\sum_{c \in C_{m,n}} C_{m,n}(c) = m^n. \quad (5.26)$$

The coefficients in the second summand of the first line in (5.25) are subject to the following recurrence relations

$$w_1(m,n+1) = m^n + (m+1)w_1(m,n), \quad (5.27)$$

$$w_k(m,n+1) = w_{k-1}(m,n) + (m+k)w_k(m,n), \quad k = 2, \ldots, n,$$

$$w_{n+1}(m,n+1) = w_n(m,n) = 1,$$

that can be deduced in the same way as we obtained the estimate in (5.20). In the first line of (5.27) we take into account also (5.26). The initial condition $w_1(m,1) = 1$ can easily be derived from (5.18). Then iterating back to $n = 1$ in the first line of (5.27) yields $w_1(m,n) = (m+1)^n - m^n$. It turns out that the complete solution of (5.27) has the following simple form

$$w_k(m,n) = \Delta^k m^n = \frac{1}{k!} \sum_{s=0}^{k} \binom{k}{s} (-1)^{k-s} (m+s)^n, \quad (5.28)$$

where $\Delta$ is the forward difference operator – a standard combinatorial object. Note that the right-hand side of (5.28) makes sense for all $k \in \mathbb{N}_0$: $w_0(m,n) = m^n$, $w_k(m,n) = 0$ for all $k > n$.

In view of (5.23) and Proposition 2.15, all the terms of the linear combination in the first line in (5.25) are continuous bounded functions of $\gamma$. Hence, the same is $\Phi_{\tau,n}^m$. However, its bound may depend on $n$, and our aim now is to control this dependence. For $\rho > 0$, set

$$T_{\tau,\rho}^m(\gamma) = \sum_{n=0}^{+\infty} \frac{\rho^n}{n!} \Phi_{\tau,n}^m(\gamma), \quad \tau \in (0,1]. \quad (5.29)$$

To get an upper bound for $T_{\tau,\rho}^m$, we estimate each $\theta^q$ in the first line of (5.25) as $\theta^q \leq c_q^q \psi$, $q \geq 0$, see (5.21), which by (5.23) and (5.22) yields

$$V_{\tau}(c; \gamma) \leq c_q^{q_1 + \cdots + q_m} \hat{F}_{\tau}^m(\gamma) = c_q^m \hat{F}_{\tau}^m(\gamma),$$

where we have taken into account that $q_1 + \cdots + q_m = c_1 + 2c_2 + \cdots + kc_k + \cdots = n$. In
A Markov process for a particle system

view of (5.26), this leads to the following

\[
\mathcal{Y}^m_{\tau, \rho}(\gamma) \leq \sum_{n=0}^{+\infty} \left( \frac{c_a \rho}{n!} \right)^n \tau^n w_k(m, n) \mathcal{F}^m_k(\gamma) \\
= \sum_{k=0}^{+\infty} \tau^k \left( \sum_{n=0}^{+\infty} \left( \frac{c_a \rho}{n!} \right)^n w_k(m, n) \right) \mathcal{F}^m_k(\gamma) \\
= \sum_{k=0}^{+\infty} \tau^k \left( \sum_{n=0}^{+\infty} \left( \frac{c_a \rho}{n!} \right)^n w_k(m, n) \right) \mathcal{F}^m_k(\gamma) \\
= \sum_{k=0}^{+\infty} \frac{\tau k}{k!} \sum_{s=0}^{k} \left( \frac{k}{s} \right) (-1)^k \sigma \left( \sum_{n=0}^{+\infty} \left( \frac{c_a \rho (m + s)}{n!} \right) \mathcal{F}^m_k(\gamma) \right) \\
= e^{c_a \rho m} \sum_{k=0}^{+\infty} \frac{\tau k}{k!} (e^{c_a \rho} - 1)^k \mathcal{F}^m_k(\gamma). \tag{5.30}
\]

Here we used the fact that \( \Delta^k m^n = 0 \) for \( k > n \), see (5.28). To proceed further we use Proposition 2.15 and (3.13) and then obtain

\[
\mathcal{F}^m_k(\gamma) \leq e^{(m+k) \Psi^m_k(\gamma)} \exp \left(-\tau \Psi^m_0(\gamma) \right),
\]

which in turn yields the last line of (5.30) the following estimate

\[
\mathcal{Y}^m_{\tau, \rho}(\gamma) \leq e^{m(c_a \rho + \tau)} \Psi^m_0(\gamma) \exp \left( -\tau \Psi^m_0(\gamma) \left( 1 - e^{c_a \rho} - 1 \right) \right) \\
\leq e^{m(c_a \rho + \tau)} \Psi^m_0(\gamma) \exp \left( -\tau \Psi^m_0(\gamma) \right) \\
\leq \left( \frac{m}{e^{c_a \rho}} \right)^m (e + 1 - \varepsilon)^m =: \delta_m(\tau),
\]

holding for some fixed \( \varepsilon \in (0, 1) \) and all

\[
\rho \leq \rho_\varepsilon := \frac{1}{c_a} \left[ \log(1 + e - \varepsilon) - 1 \right]. \tag{5.31}
\]

By (5.29) this yields the estimate in question in the following form

\[
\Phi^m_{\tau, \rho}(\gamma) \leq \frac{n!}{\rho c_a} \delta_m(\tau), \quad \tau \in (0, 1]. \tag{5.32}
\]

5.3 Proof of Lemma 5.1

According to Definition 3.1 and (5.17), we have that \( \Phi^m_{\tau} \) lies in the linear span of \( \mathcal{F} \) for each \( \tau > 0 \) and \( m \in \mathbb{N} \). If \( \{\mu_t\}_{t \geq 0} \subset \mathcal{P}(\Gamma_\tau) \) solves (3.18), then

\[
\mu_t(\Phi^m_{\tau,t}) = \mu_0(\Phi^m_{\tau,t}) + \int_0^t \mu_u(L \Phi^m_{\tau,t}) du \leq \mu_0(\Phi^m_{\tau,t}) + \int_0^t \mu_u(\Phi^m_{\tau,t}) du, \tag{5.33}
\]

where we have used (5.18). Since \( \Phi^m_{\tau,1} \) is a linear combination of the elements of \( \mathcal{F} \), we can repeat (5.33) with this function and obtain

\[
\mu_t(\Phi^m_{\tau,1}) \leq \mu_0(\Phi^m_{\tau,1}) + \int_0^t \mu_u(\Phi^m_{\tau,1}) du.
\]
which then can be used in (5.33). In view of (5.24), we can repeat this procedure due
times and thereby get the following estimate
\[
\mu_t(\Phi^m_t) \leq \sum_{k=0}^{n-1} \frac{t^k}{k!} \mu_0(\Phi^m_{\tau,k}) + \int_0^t \int_0^{t_1} \cdots \int_0^{t_{n-1}} \mu_{n}(\Phi^m_{\tau,n}) dt_n dt_{n-1} \cdots dt_1 \tag{5.34}
\]
\[
\leq \sum_{k=0}^{n-1} \frac{t^k}{k!} \mu_0(\Phi^m_{\tau,k}) + \left( \frac{1}{\rho_\tau} \right)^n \delta_m(\tau),
\]
where we have used (5.32) and the fact that \(\mu_1\) is a probability measure. For \(t < \rho_\tau\), the
last summand in the right-hand side of (5.34) vanishes as \(n \to +\infty\). Hence,
\[
\mu_t(\Phi^m_t) \leq \sum_{n=0}^{+\infty} \frac{n^t}{n!} \mu_0(\Phi^m_{\tau,n}), \quad t < \rho_\tau, \quad \tau \in (0, 1].
\tag{5.35}
\]
By (2.42) and (5.23) it follows that the element of \(\tilde{F}\) in the first summand in the first line
in (5.25) satisfies
\[
V_\tau(c; \gamma) \leq V_0(c; \gamma) := \sum_{x_1 \in \gamma} \theta^{q_1}(x_1) \sum_{x_2 \in \gamma \setminus x_1} \theta^{q_2}(x_2) \cdots \sum_{x_m \in \gamma \setminus \{x_1, \ldots, x_{m-1}\}} \theta^{q_n}(x_m), \quad \theta \in \Theta^+.
\]
\(V_0(c; \cdot)\) is an unbounded function, which, however, is \(\mu_0\)-integrable. Let \(x_0\) be the type of
\(\mu_0\). As in Remark 2.5, we then have
\[
\mu_0(V_\tau(c; \cdot)) \leq \pi_{\varphi_0}(V_0(c; \cdot)) = \chi_0^m(\theta^{q_1}) \cdots \langle \theta^{q_n} \rangle = 2^n \langle \varphi_0(\theta) \rangle^m,
\tag{5.36}
\]
where \(\langle \theta^{q_j} \rangle := \int_{R^d} \theta^{q_j}(x) dx = 2^{q_j} \int_{R^d} \theta(x) dx = 2^{q_j} \langle \theta \rangle\),
see (5.21) and (3.2). Here we have taken into account that \(q_1 + \cdots + q_m = n\). By (3.13)
we have
\[
\tilde{F}_{\tau+k}^m(\gamma) \leq \sum_{x_1 \in \gamma} \psi(x_1) \sum_{x_2 \in \gamma \setminus x_1} \psi(x_2) \cdots \sum_{x_{m+k} \in \gamma \setminus \{x_1, \ldots, x_{m+k-1}\}} \psi(x_{m+k}).
\]
Then similarly as in (5.36) we obtain
\[
\mu_0(\tilde{F}_{\tau+k}^m) \leq (\varphi_0(\psi))^m = \langle \varphi_0(\psi) \rangle^m.
\tag{5.37}
\]
We use (5.36) and (5.37) in (5.25) and then in (5.35) and arrive at the following estimate
\[
\mu_t(\Phi^m_t) \leq (e^{2t} \varphi_0(\theta))^m + (\varphi_0(\psi))^m e^{ctm} \sum_{k=1}^{\infty} \frac{t^k}{k!} \langle \varphi_0(\psi) \rangle^k (e^{ct} - 1)^k
\]
\[
\leq (e^{2t} \varphi_0(\theta))^m + \tau (\varphi_0(\psi))^m e^{ctm} (e^{ct} - 1) \exp (\varphi_0(\psi) (e^{ct} - 1)),
\]
where we have applied the same approach as in obtaining (5.30) and the fact that \(\tau \leq 1\).
Since, for each \(\gamma \in \Gamma_1\), and an arbitrary sequence \(\tau_n \to 0\), \{\tilde{F}_w^0(\gamma)\}_{n \in \mathbb{N}}\ is a nondecreasing
sequence, by (5.17) and Beppo Levi’s monotone convergence theorem we then get from
the latter that, cf. (5.16),
\[
\lim_{\tau \to 0} \mu_t(\Phi^m_{\tau}) = \mu_t(F^m_t) = \langle \mu^{m}_{\tau}, \theta^{\otimes m} \rangle \tag{5.38}
\]
\[
:= \int_{R^d} \mu^{m}_{\tau}(x_1, \ldots, x_m) \theta(x_1) \cdots \theta(x_m) dx_1 \cdots dx_m
\]
\[
\leq (e^{2t} \varphi_0(\theta))^m,
\]
The set of functions \( \Theta^+ \) defined in (2.32) is closed with respect to the pointwise multiplication and separates points of \( \mathbb{R}^d \). Such functions vanish at infinity and are everywhere positive. Then by the aforementioned version of the Stone-Weierstrass theorem [10] the linear span of this set is dense (in the supremum norm) in the algebra \( C_0(\mathbb{R}^d) \) of all continuous functions that vanish at infinity. At the same time, \( C_0(\mathbb{R}^d) \cap L^1(\mathbb{R}^d) \) is dense in \( L^1(\mathbb{R}^d) \). Therefore, by (5.39) the maps \( \theta \mapsto \langle \mu^{(m)}_\theta \circ \varphi^m \rangle, m \in \mathbb{N} \) can be extended to homogeneous continuous monomials on \( L^1(\mathbb{R}^d) \). This yields the proof of the considered statement for \( t < \rho_z \), see Remark 2.4. Since \( \rho_z \) is independent of \( z_0 \), the continuation to all \( t > 0 \) can be made by the repetition of the same arguments.

### 5.4 Proof of the uniqueness

By employing Lemma 5.1 and Corollary 5.3, see also Remark 3.5, we prove the following statement.

**Lemma 5.4.** Assume that two solutions \( \{P^{(i)}_{s,\mu} : s \geq 0, \mu \in \mathcal{P}_{\text{exp}}\} \), \( i = 1, 2 \), see Definition 3.3, satisfy \( P^{(i)}_{s,\mu} \circ \varphi_t^{-1} = P^{(i)}_{s,\mu} \circ \varphi_t^{-1} \) for all \( t \geq s, s \geq 0 \) and \( \mu \in \mathcal{P}_{\text{exp}} \). Then \( P^{(i)}_{s,\mu} = P^{(i)}_{s,\mu} \) for all \( s \) and \( \mu \).

**Proof.** By Kolmogorov’s extension theorem it is enough to prove that all finite-dimensional marginals of both path measures coincide. In view of claim (i) of Proposition 2.14, to this end we have to show that the following holds

\[
P^{(1)}_{s,\mu} (F_{t_1} \cdots F_{t_n}) = P^{(2)}_{s,\mu} (F_{t_1} \cdots F_{t_n}),
\]

where \( F_{t_i} (\gamma) = \tilde{\varphi}_{t_i}^\theta (\varphi_{t_i} (\gamma)) \), \( i = 1, \ldots, n \), see (3.4), ought to be taken with all possible \( \theta_i \in \Theta^+_\psi, \tau_i > c \theta_i \) and \( t_i \) satisfying \( s \leq t_1 \leq \cdots \leq t_n \). Assume that (5.40) holds with a given \( n \) and prove its validity for \( n + 1 \). Since \( F_{t_i} (\gamma) > 0 \), see (2.36), we may set

\[
C^{-1} = P^{(1)}_{s,\mu} (F_{t_1} \cdots F_{t_n}),
\]

and then define two path measures on \((\mathcal{D}_{[t_n, +\infty)}, \mathfrak{N}_{t_n, +\infty})\)

\[
Q^i (B) = CP^{(i)}_{s,\mu} (F_{t_1} \cdots F_{t_n} 1_B), \quad i = 1, 2.
\]

Since both \( P^i \) satisfy (3.16), we have also

\[
\int_{\mathcal{D}_{[t_n, +\infty)}} H(\gamma) Q^i (d\gamma) = 0, \quad i = 1, 2.
\]

Hence, both maps \([t_n, +\infty) \ni t \mapsto Q^i \circ \varphi_t^{-1} =: \mu^i_t \in \mathcal{P}(\Gamma_s), i = 1, 2 \) solve

\[
\mu^1_{u_2} (F) = \mu^1_{u_1} (F) + \int_{u_1}^{u_2} \mu^1_v (\text{L} F) dv, \quad F \in \mathcal{D}(\text{L}),
\]

for all \( u_2 > u_1 \geq t_n \), see Remark 3.5. By the inductive assumption and claim (iv) of Proposition 3.2 it follows that \( \mu^{(i)}_{t_n} = \mu^{(i)}_{t_n} =: \mu \in \mathcal{P}_{\text{exp}} \). By Lemma 5.1 we then conclude that \( \mu^1_t \in \mathcal{P}_{\text{exp}}, i = 1, 2 \) for all \( t > t_n \). That is, both \( Q^i \) satisfy all the three conditions of Definition 3.3 and thus belong to solutions of the restricted initial value martingale problem. Hence, \( \mu^1_t = \mu^2_t \) by the assumption of the lemma. In particular,

\[
\mu^{(2)}_{t_{n+1}} (\tilde{\varphi}^{\theta_{t_{n+1}}} = \mu^{(2)}_{t_{n+1}} (\tilde{\varphi}^{\theta_{t_{n+1}}}),
\]

which completes the proof.
A Markov process for a particle system

**Theorem 5.5.** Let \( \{P_{s,\mu}^{(i)} : s \geq 0, \mu \in \mathcal{P}_{\exp}\}, i = 1, 2 \) be two solution of the restricted initial value martingale problem in the sense of Definition 3.3. Then \( P_{s,\mu}^{(1)} = P_{s,\mu}^{(2)} \) for all \( s \geq 0 \) and \( \mu \in \mathcal{P}_{\exp} \).

**Proof.** By Remark 3.5 both \( P_{s,\mu}^{(1)} \circ \varpi_t, t \geq s \) solve (3.18), which by Theorem 5.2 yields \( P_{s,\mu}^{(1)} \circ \varpi_t^{-1} = P_{s,\mu}^{(2)} \circ \varpi_t^{-1} \), holding for all \( t \geq s \) and \( \mu \in \mathcal{P}_{\exp} \). Then the proof follows by Lemma 5.4.

---

### 6 The Existence: Approximating Models

The aim of this and the subsequent sections is to prove the following statement which is the second cornerstone in the proof of Theorem 3.6.

**Theorem 6.1.** There exists a family of probability measures which solves the restricted initial value martingale problem for our model in the sense of Definition 3.3.

The basic idea is to approximate the model by auxiliary models described by \( L^\alpha \), \( \alpha \in [0, 1] \) with \( L^0 \) coinciding with \( L \) defined in (1.2). For \( \alpha \in (0, 1] \), the solution \( \{P_{s,\mu}^{\alpha} : s \geq 0, \mu \in \mathcal{P}_{\exp}\} \) of the corresponding restricted initial value martingale problem for \( L^\alpha \) will be constructed in a direct way. Then the proof of Theorem 6.1 will be done by showing the weak convergence \( P_{s,\mu}^{\alpha} \Rightarrow P_{s,\mu} \) as \( \alpha \to 0 \), and then by proving that \( \{P_{s,\mu} : s \geq 0, \mu \in \mathcal{P}_{\exp}\} \) is a solution in question. In the current section, we introduce the auxiliary models and study their relations with the basic model. The construction of the path measures \( P_{s,\mu}^{\alpha} \) will be performed in the subsequent section.

#### 6.1 The approximating models

Recall that \( \psi \) was introduced in (2.17), see also (2.14). Along with these functions, we shall use \( \Psi_1(\gamma) = 1 + \Psi(\gamma) \) and

\[
\psi_\alpha(x) = \frac{1}{1 + \alpha |x|^{d+1}}, \quad \alpha \in [0, 1].
\]

Set

\[
a_\alpha(x, y) = a(x - y)\psi_\alpha(x), \quad x, y \in \mathbb{R}^d.
\]

Note that \( a_0(x, y) = a(x - y) \) and \( a_\alpha(x, y) \neq a_\alpha(y, x) \) for \( \alpha \in (0, 1] \). Now let \( L^\alpha \) be defined as in (1.2) with \( a \) replaced by \( a_\alpha \). That is,

\[
(L^\alpha F)(\gamma) = \int_{\mathbb{R}^d} \psi_\alpha(x) a(x - y) \exp \left( -\sum_{z \in \gamma \setminus x} \phi(z - y) \right) \left[ F(\gamma \setminus x \cup y) - F(\gamma) \right] dy.
\]

Then keeping in mind (4.2) and (4.7) we define \( L^{\Delta \alpha} \) by the following expression

\[
\mu(L^{\alpha} F^\theta) = \langle L^{\Delta \alpha} k_{\mu_e}, e(\theta, \cdot) \rangle, \quad \alpha \in [0, 1].
\]

One observes that \( L^{\Delta 0} \) coincides with the operator introduced in (4.8). For \( \alpha \in (0, 1] \), \( L^{\Delta \alpha} \) is then obtained by replacing in (4.8) \( a(x - y) \) by \( a_\alpha(x, y) \leq a(x - y) \). Hence, \( L^{\Delta \alpha} \) clearly satisfies (4.13) and similar estimates. Then by repeating the construction realized in subsection 4.2 we obtain the family of bounded operators \( \{Q_{\theta, \theta'}(t) : t \in [0, T(\theta', \theta)]\} \) (resp. \( \{H_{\theta, \theta'}(t) : t \in [0, T(\theta', \theta)]\} \) \( \theta' > \theta \) acting from \( \mathcal{K}_\theta \) to \( \mathcal{K}_{\theta'} \) (resp. from \( \mathcal{G}_{\theta'} \) to \( \mathcal{G}_\theta \)). By employing these families we then set

\[
k^\alpha_t = Q^\alpha_{\theta_t}(t)k_0, \quad G^\alpha_t = H^\alpha_{\theta_t}(t)G_0,
\]

with \( k_0 \in \mathcal{K}_\theta \) and \( G_0 \in \mathcal{G}_{\theta'} \). Note that, for \( \alpha = 0 \), these vectors coincide with those introduced in (4.21) and (4.22), respectively, and thus they satisfy (4.23) for all \( \alpha \in [0, 1] \).
Moreover, as in Proposition 4.2, for each \( \vartheta_0 \in \mathbb{R} \) and \( \mu \in \mathcal{P}_\text{exp}^{\vartheta_0} \), by (6.4) with \( k_0 = k_\mu \) we obtain a family, \( \{ \mu_t^\omega : t \geq 0, \mu_0 = \mu \} \subset \mathcal{P}_\text{exp}, \mu_t^\omega \in \mathcal{P}_\text{exp}^{\vartheta_0} \) such that

\[
\mu_t^\omega (F^\vartheta) = \langle \langle k_t^\omega, \epsilon(\theta, \cdot) \rangle \rangle, \quad \theta \in L^1(\mathbb{R}^d). \tag{6.5}
\]

Next, by repeating the construction used in the proof of Theorem 5.2 one obtains that the map \( t \mapsto \mu_t^\omega \) is a unique solution of the equation

\[
\mu_t^\omega (F) = \mu_t^\omega (F) + \int_{t_1}^{t_2} \mu_u^\omega (L^\omega F) du, \quad t_2 > t_1 \geq 0,
\]

holding for all \( F : \Gamma_\ast \to \mathbb{R} \) which can be written as \( F = KG \) with \( G \in \cap_{\vartheta \in \mathbb{R}} \mathcal{G}_\vartheta \), see Corollary 5.3. Here and below we set

\[
\mathcal{D}(L^\omega) = \mathcal{D}(L), \quad \alpha \in (0, 1],
\]

with \( \mathcal{D}(L) \) as in Definition 3.1.

### 6.2 The weak convergence

Our aim now is to prove that the families \( \{ \mu_t^\omega : t \geq 0, \mu_0 = \mu \} \subset \mathcal{P}_\text{exp}, \alpha \in [0, 1] \) constructed above have the following property.

**Lemma 6.2.** For each \( t > 0 \), it follows that \( \mu_t^\omega \rightharpoonup \mu_t \) as \( \alpha \to 0 \), where we mean the weak convergence of measures on the Polish space \( \Gamma_\ast \).

We begin by proving the convergence of the corresponding correlation functions.

**Lemma 6.3.** For each \( t > 0 \), one finds \( \bar{\vartheta}_t > \vartheta_t \) such that the following holds

\[
\forall G \in \mathcal{G}_{\bar{\vartheta}_t}, \quad \langle \langle k_t^\omega, G \rangle \rangle \to \langle \langle k_t, G \rangle \rangle, \quad \text{as } \alpha \to 0. \tag{6.6}
\]

**Proof.** We recall that \( k_t \) satisfies (4.26) with \( L_{\vartheta}^\omega \) corresponding to \( \alpha = 0 \). Note that the domains of \( L_{\vartheta}^\Delta \) are the same for all \( \alpha \in [0, 1] \).

Assume now that the convergence stated in (6.6) holds for a given \( t \geq 0 \). Note that \( k_0 = k_0^\omega = k_\mu \); hence, this assumption is valid for at least \( t = 0 \). Let us prove that there exists \( s_0 > 0 \) possibly dependent on \( t \) such that this convergence holds for all \( t + s, s \leq s_0 \). Keeping in mind that \( Q^\omega \) and \( k_t^\omega \) satisfy the corresponding analogs of (4.20) and (4.26), respectively, we write

\[
k_{t+s} - k_{t+s}^\omega = Q_{\bar{\vartheta}_t, \vartheta_t}(s) k_t - Q_{\bar{\vartheta}_t, \vartheta_t}(s) k_t^\omega, \tag{6.7}
\]

where \( \bar{\vartheta}_t = t_0 + \vartheta_t \) and \( \vartheta_t = \vartheta_0 + t \). Note that the left-hand side of (6.7) is considered as a vector in \( \mathcal{K}_{\bar{\vartheta}_t} \). Both \( Q_{\bar{\vartheta}_t, \vartheta_t}(s) \) and \( Q_{\bar{\vartheta}_t, \vartheta_t}^\omega(s) \) are defined only for \( s < \tau(\bar{\vartheta}_t) \), see (4.15). At the same time, for each \( \vartheta' > \vartheta_t \), \( Q_{\vartheta', \vartheta_t}(0) = Q_{\vartheta', \vartheta_t}^\omega(0) = I_{\vartheta'} \), where the latter is the embedding operator, see (4.6). Keeping this (4.20) and (4.26) in mind we rewrite (6.7) as follows

\[
k_{t+s} - k_{t+s}^\omega = Q_{\bar{\vartheta}_t, \vartheta_t}(s)(k_t - k_t^\omega) - \left( \int_0^s \frac{d}{ds} Q_{\bar{\vartheta}_t, \vartheta_t}(s-u) Q_{\bar{\vartheta}_t, \vartheta_t}(u) du \right) k_t^\omega \tag{6.8}
\]

\[
= Q_{\bar{\vartheta}_t, \vartheta_t}(s)(k_t - k_t^\omega) + \int_0^s Q_{\bar{\vartheta}_t, \vartheta_t}(s-u)L_{\vartheta' \vartheta_t}^\omega Q_{\bar{\vartheta}_t, \vartheta_t}(u) k_t^\omega du
\]

\[
- \int_0^s Q_{\bar{\vartheta}_t, \vartheta_t}(s-u)L_{\vartheta' \vartheta_t}^\omega Q_{\bar{\vartheta}_t, \vartheta_t}(u) k_t^\omega du
\]

\[
= Q_{\bar{\vartheta}_t, \vartheta_t}(s)(k_t - k_t^\omega) + \int_0^s Q_{\bar{\vartheta}_t, \vartheta_t}(s-u)L_{\vartheta' \vartheta_t}^\omega k_{t+u}^\omega du,
\]
where \( \tilde{L}^{\Delta, \alpha} \) is given in (4.8) with \( a(x - y) \) replaced by \( \tilde{a}_\alpha(x, y) = a(x - y)(1 - \psi_\alpha(x)) \). The choice of \( s \) and \( \vartheta_1, \vartheta_2 \) should be made in such a way that the series as in (4.18) converge for the corresponding operators. Set \( \vartheta_1 = \vartheta + \delta(\vartheta_t)/2 \). We use this in (4.14) and obtain that

\[
T(\tilde{\vartheta}_t, \vartheta_1) = \frac{\tau(\vartheta_t)}{2} < T(\vartheta_1, \vartheta_t). \tag{6.9}
\]

Then for some \( \epsilon \in (0, 1) \), we set

\[
s_0 = \epsilon(\vartheta_t)/2 = \epsilon T(\tilde{\vartheta}_t, \vartheta_1). \tag{6.10}
\]

Since the map \( \vartheta \mapsto T(\tilde{\vartheta}_t, \vartheta) \) is continuous, one can find \( \vartheta_2 \in (\vartheta_1, \vartheta_t) \) such that \( s_0 < T(\vartheta_1, \vartheta_2) \), cf. (6.10), which together with (6.9) yields that all the three \( Q_{\vartheta_1, \vartheta_2}(s - u) \), \( Q_{\vartheta_2, \vartheta}(s - u) \) and \( Q^9_{\vartheta_1, \vartheta}(s - u) \) in (6.8) are defined for all \( s \leq s_0 \) and \( u \in [0, s] \). Now we take \( G \in \mathcal{G}_{\vartheta_2} \), and set \( G_s = H_{\vartheta_2, \vartheta}(s) G \), \( s \leq s_0 \). Then \( G_s \in \mathcal{G}_{\vartheta_2} \subset \mathcal{G}_{\vartheta_1} \), which yields by (6.8) the following

\[
\langle \langle k_{t+s} - k_{t+s}^\alpha, G_s \rangle \rangle = \langle \langle k_t - k_t^\alpha, G_s \rangle \rangle + Y_\alpha(s), \tag{6.11}
\]

\[
Y_\alpha(s) := \int_0^s \langle \langle \tilde{L}^{\Delta, \alpha} k_t^\alpha, G_{s-u} \rangle \rangle du.
\]

Thus, we have to prove that \( Y_\alpha(s) \to 0 \) as \( \alpha \to 0 \). Since \( L^{\Delta, \alpha} \) consists of two terms, see (4.8), it is convenient for us to write \( Y_\alpha(s) = Y_\alpha^{(1)}(s) + Y_\alpha^{(2)}(s) \), where

\[
Y_\alpha^{(1)}(s) = \int_0^s \int_{\mathbb{R}^d} \left( \sum_{y \in \mathbb{N}} \int_{\mathbb{R}^d} \tilde{a}_\alpha(x, y)e(\tau_y; \eta \setminus y)(W_y k_{t+u}^\alpha)(\eta \setminus y \cup x) dx \right) \times G_{s-u}(\eta) \lambda(d\eta) du \tag{6.12}
\]

and

\[
Y_\alpha^{(2)}(s) = -\int_0^s \int_{\mathbb{R}^d} \left( \sum_{x \in \mathbb{N}} \int_{\mathbb{R}^d} \tilde{a}_\alpha(x, y)e(\tau_y; \eta \setminus x)(W_y k_{t+u}^\alpha)(\eta \setminus x) G_{s-u}(\eta \cup y) dx dy \right) \lambda(d\eta) du, \tag{6.13}
\]

To estimate both terms we take into account that \( e(\tau_y; \eta) \leq 1 \) and

\[
|\langle W_y k_{t+u}^\alpha(\eta \cup x) \rangle| \leq \exp(\vartheta_1 + \vartheta_1|\eta| + \langle \vartheta \rangle e^\vartheta),
\]

where the latter estimate follows by the fact that \( k_{t+u}^\alpha(\eta) \leq \exp(\vartheta_1|\eta|) \leq \exp(\vartheta_1|\eta|) \), see claim (a) of Proposition 4.2. By these estimates we obtain from (6.12) and (6.13) the following

\[
\left| Y_\alpha^{(i)}(s) \right| \leq \int_{\mathbb{R}^d} h^{(i)}_\alpha(y) g^{(i)}(y) dy, \quad i = 1, 2, \tag{6.14}
\]

where

\[
h^{(1)}_\alpha(y) = \int_{\mathbb{R}^d} \tilde{a}_\alpha(x, y) dx = \int_{\mathbb{R}^d} (1 - \psi_\alpha(|x|)) a(x - y) dx, \tag{6.15}
\]
where the second term of the last line was obtained by Markov’s inequality and (3.3) yields the proof of (6.6) for $g$ and $Y$ theorem and (6.16) and (6.14) we conclude that $Y(\varphi)$. Hence, for each $t$, together with the estimate
\[ r_1 := \exp(\varphi + \langle \varphi \rangle^2) \]
Let us show that $g(1) = g(1)$ is integrable for all $s \leq s_0$. To this end we use the fact that $G_{s-u} \in G_\varphi$ for all $s \leq s_0$ and $u \leq s$. Then its norm can be estimated
\[ \|G_{s-u}\|_\varphi \leq \frac{T(\varphi_1, \varphi_2)}{T(\varphi_1, \varphi_2) - s_0} |G|_\varphi := C_G \]
which is finite by our choice of $\varphi_2$ and $s_0$. Then
\begin{align}
\int_{\mathbb{R}^d} g_s(1)(y) dy & = c(\varphi_1) \int_0^s \int_{\mathbb{R}^d} |G_{s-u}(\eta \cup y)| e^{\varphi_1 |\eta|} |dy\lambda(\eta)| du \\
& = c(\varphi_1) e^{-\varphi_1} \int_0^s \int_{\mathbb{R}^d} |G_{s-u}(\eta)| e^{\varphi_1 |\eta|} |dy\lambda(\eta)| du \\
& = c(\varphi_1) e^{-\varphi_1} \int_0^s \int_{\mathbb{R}^d} |\eta| e^{\varphi_1 |\eta|} |dy\lambda(\eta)| du \\
& \leq \frac{c(\varphi_1)s}{e^{1+\varphi_1}(\varphi_2 - \varphi_1)} C_G.
\end{align}

Now let us turn to (6.15). First of all, we note that $h_\alpha(1) \leq 1$, see (3.2). The function $r \mapsto 1 - \tilde{\psi}_\alpha(r)$ is increasing. Then, for a certain $r > 0$, we have
\begin{align}
\tilde{h}_\alpha(1) & = \int_{\mathbb{R}^d} \left(1 - \tilde{\psi}_\alpha(|x + y|)\right) a(x) dx \\
& \leq \int_{B_r} \left(1 - \tilde{\psi}_\alpha(|x + y|)\right) a(x) dx + \int_{B_r^c} a(x) dx \\
& \leq (1 - \tilde{\psi}_\alpha(|y|)) + \frac{m_{\alpha+1}^{d+1}}{r^{d+1}},
\end{align}
where the second term of the last line was obtained by Markov’s inequality and (3.3) together with the estimate $1 - \tilde{\psi}_\alpha(r) \leq 1$. Now we set in (6.17) $r = \alpha^{-1/(d+2)}$ and obtain
\[ h_\alpha(1) \leq \frac{\alpha^{1/(d+2)}(1 + |y|)^{d+1}}{1 + \alpha^{2/(d+2)}(1 + |y|)^{d+1}} + m_{\alpha+1}^{d+1} \frac{d+1}{r^{d+1}}.
\]
Hence, for each $y$, $h_\alpha(1) \to 0$ as $\alpha \to 0$. Then by Lebesgue’s dominated convergence theorem and (6.16) and (6.14) we conclude that $Y_\alpha(1)(s) \to 0$ as $\alpha \to 0$, holding for all $s \leq s_0$.

Now we turn to (6.13) by which we get
\[ h_\alpha(1)(y) = \int_{\mathbb{R}^d} \tilde{\psi}_\alpha(y) a(x - y) dx = \tilde{\psi}_\alpha(y) := 1 - \psi_\alpha(y), \]
and $g_\alpha(1) = g_\alpha(1)$. Hence, also $Y_\alpha(1)(s) \to 0$ as $\alpha \to 0$, holding for all $s \leq s_0$, which by (6.11) yields the proof of (6.6) for $t + s$ with $s \leq s_0$ whenever it holds for $t$. To complete the proof let us consider the following sequences, cf. (6.10),
\begin{align}
t_l = t_{l-1} + s_{0l}, & \quad t_0 = 0, \quad l \in \mathbb{N}, \\
s_{0l} = \epsilon \tau(\varphi_{t_{l-1}})/2.
\end{align}
A Markov process for a particle system

Since \( k_0^\alpha = k_0 = k_\mu \), the proof made above yields the stated convergence for \( t \leq \sup_l t_l = \lim_l t_l \). Thus, our aim is to show that \( t_l \to +\infty \) as \( l \to +\infty \). Assume that \( \sup_l t_l = t_* < \infty \).

By the first line in (6.18) we have that \( t_l = s_{t_l} + \cdots + s_{t_0} \) and hence \( s_{t_0} \to 0 \) in this case. Now we pass in the second line of (6.18) to the limit \( l \to +\infty \) (\( \tau \) is continuous) and get that \( t_* \) should satisfy \( \tau(\theta_{t_*}) = \tau(\theta_0 + t_*) = 0 \), which is impossible as \( \tau(\theta) > 0 \) for all \( \theta \in \mathbb{R} \). This completes the proof with \( \tilde{\theta} = \theta_{t_*} \).

\[
\text{Proof of Lemma 6.2.} \quad \text{By Lemma 6.3 and (2.7) it follows that} \quad \mu^\alpha_t(F) \to \mu_t(F) \quad \text{as} \quad \alpha \to 0, \quad \text{holding for all} \quad F \in \mathcal{F}, \quad \text{see (5.1). Then the proof follows by the fact that} \quad \tilde{\mathcal{F}} \subset \mathcal{F}, \quad \text{see (5.2), and claim (ii) of Proposition 2.14.} \quad \square
\]

Below we use the following fact, that can be considered as a complement to Lemma 6.2.

\[\text{Lemma 6.4.} \quad \text{Assume that a sequence} \quad \{\nu_n\}_{n \in \mathbb{N}} \subset \mathcal{P}_{\exp}^\alpha, \quad \theta \in \mathbb{R}, \quad \text{cf. (4.25), satisfy} \quad \nu_n \to \nu \quad \text{as} \quad n \to +\infty \quad \text{for some} \quad \nu \in \mathcal{P}(\Gamma_\ast). \quad \text{Then} \quad \nu \in \mathcal{P}_{\exp}^\alpha. \quad \text{Furthermore, for each} \quad G \in \cap_\theta \mathcal{G}_\theta, \quad \text{it follows that}
\]

\[
\langle \langle k_{\nu_n}, G \rangle \rangle \to \langle \langle k_\nu, G \rangle \rangle, \quad n \to +\infty. \tag{6.19}
\]

\[\text{Proof.} \quad \text{By assumption,} \quad \nu_n(F) \to \nu(F) \quad \text{for each} \quad F \in \tilde{\mathcal{F}}, \quad \text{see (3.4) and Proposition 2.15. By (2.42), (5.17) and (5.16), for given} \quad m \in \mathbb{N}, \quad \theta \in \Theta_{\mathcal{E}}^\alpha \quad \text{and} \quad \tau \in (0, 1], \quad \text{we then get}
\]

\[
\nu(\Phi^m_{\tau}) \leq \sup_{n \in \mathbb{N}} \nu_n(\Phi^m_{\tau}) \leq e^{md} \| \theta \|^m.
\]

Then the proof of \( \nu \in \mathcal{P}_{\exp}^\alpha \) follows by the monotone convergence theorem and (2.8). The validity of (6.19) for \( G \) such that \( KG \in \tilde{\mathcal{F}} \) follows by the fact just mentioned, i.e., just because \( \nu \) has a correlation function. The extension of (6.19) to all \( G \in \cap_\theta \mathcal{G}_\theta \) is then made by the same arguments as the proof of (5.11). \( \square \)

7 The Existence: Approximating Processes

In this section, we prove Theorem 6.1 by constructing path measures for the models described by \( L^\alpha \), \( \alpha \in (0, 1] \) introduced in the preceding section. This will be done in a direct way by means of the corresponding Markov transition functions.

7.1 The Markov transition functions

The transition functions in question will be obtained in the form

\[
p^\alpha_t(\gamma, \cdot) = S^\alpha(t)\delta_\gamma, \quad t \geq 0, \quad \alpha \in (0, 1], \tag{7.1}
\]

where \( \delta_\gamma \) is the Dirac measure with atom at \( \gamma \in \Gamma_\ast \), and \( S^\alpha = \{ S^\alpha(t) \}_{t \geq 0} \) is a stochastic semigroup of linear operators, related to the Kolmogorov operator \( L^\alpha \). Hence, we begin by constructing \( S^\alpha \).

7.1.1 Stochastic semigroups

A more detailed presentation of the notions and facts which we introduce here can be found in [2, 3, 31].

Let \( \mathcal{E} \) be an ordered real Banach space, and \( \mathcal{E}^+ \) be a generating cone of its positive elements. Set \( \mathcal{E}^{+, 1} = \{ x \in \mathcal{E}^+ : \| x \|_\mathcal{E} = 1 \} \) and assume that the norm is additive on \( \mathcal{E}^+ \), i.e., \( \| x + y \|_\mathcal{E} = \| x \|_\mathcal{E} + \| y \|_\mathcal{E} \) whenever \( x, y \in \mathcal{E}^+ \). In such spaces, there exists a positive linear functional, \( \varphi_\mathcal{E} \), such that

\[
\varphi_\mathcal{E}(x) = \| x \|_\mathcal{E}, \quad x \in \mathcal{E}^+. \tag{7.2}
\]
A Markov process for a particle system

A \( C_0 \)-semigroup, \( S = \{ S(t) \}_{t \geq 0} \), of bounded linear operators on \( \mathcal{E} \) is said to be stochastic (resp. substochastic) if the following holds \( \| S(t)x \|_\mathcal{E} = 1 \) (resp. \( \| S(t)x \|_\mathcal{E} \leq 1 \)) for all \( t > 0 \) and \( x \in \mathcal{E}^{+1} \). Let \( \mathcal{D} \subset \mathcal{E} \) be a dense linear subspace, \( \mathcal{D}^+ = \mathcal{D} \cap \mathcal{E}^+ \) and \( (A, \mathcal{D}), (B, \mathcal{D}) \) be linear operators in \( \mathcal{E} \). A paramount question of the theory of stochastic semigroups is under which conditions the closure (resp. an extension) of \( (A + B, \mathcal{D}) \) is the generator of a stochastic semigroup. Classical works on this subject trace back to Feller, Kato, Miyadera, etc, see [2, 31]. In the present work, we will use a result of [31], which we present now in the form adapted to the context.

To proceed we need to further specify the properties of the space \( \mathcal{E} \).

**Assumption 7.1.** There exists a linear subspace, \( \tilde{\mathcal{E}} \subset \mathcal{E} \), which has the following properties:

(i) \( \tilde{\mathcal{E}} \) is dense in \( \mathcal{E} \) in the norm \( \| \cdot \|_\mathcal{E} \).

(ii) There exists a norm, \( \| \cdot \|_{\tilde{\mathcal{E}}} \), on \( \tilde{\mathcal{E}} \) that makes it a Banach space.

(iii) \( \tilde{\mathcal{E}}^+ := \tilde{\mathcal{E}} \cap \mathcal{E}^+ \) is a generating cone in \( \tilde{\mathcal{E}} \); \( \| \cdot \|_{\tilde{\mathcal{E}}} \) is additive on \( \tilde{\mathcal{E}}^+ \) and hence there exists a linear functional, \( \varphi_{\tilde{\mathcal{E}}} \), on \( \tilde{\mathcal{E}} \), such that \( \| x \|_{\tilde{\mathcal{E}}} = \varphi_{\tilde{\mathcal{E}}}(x) \) whenever \( x \in \tilde{\mathcal{E}}^+ \), cf. (7.2).

(iv) The cone \( \tilde{\mathcal{E}}^+ \) is dense in \( \mathcal{E}^+ \).

For \( \mathcal{D} \) as above, set \( \tilde{\mathcal{D}} = \{ x \in \mathcal{D} \cap \tilde{\mathcal{E}} : Ax \in \tilde{\mathcal{E}} \} \). Then \( (A, \tilde{\mathcal{D}}) \) is the trace of \( A \) in \( \tilde{\mathcal{E}} \). The next statement is an adaptation of [31, Theorem 2.7].

**Proposition 7.2** (Thieme-Voigt). Assume that:

(i) \( -A : \mathcal{D}^+ \to \mathcal{E}^+ \) and \( B : \mathcal{D}^+ \to \mathcal{E}^+ \);

(ii) \( (A, \mathcal{D}) \) is the generator of a substochastic semigroup, \( S = \{ S(t) \}_{t \geq 0} \), on \( \mathcal{E} \) such that \( S(t) : \mathcal{E} \to \mathcal{E} \) for all \( t \geq 0 \) and the restrictions \( S(t)|_{\tilde{\mathcal{E}}} \) constitute a \( C_0 \)-semigroup on \( \tilde{\mathcal{E}} \) generated by \( (A, \tilde{\mathcal{D}}) \);

(iii) \( B : \tilde{\mathcal{D}} \to \tilde{\mathcal{E}} \) and \( \varphi_{\tilde{\mathcal{E}}} ((A + B)x) = 0 \), for \( x \in \mathcal{D}^+ \);

(iv) there exist \( c > 0 \) and \( \varepsilon > 0 \) such that

\[
\varphi_{\tilde{\mathcal{E}}} ((A + B)x) \leq c\varphi_{\tilde{\mathcal{E}}}(x) - \varepsilon \| Ax \|_{\tilde{\mathcal{E}}} \quad \text{for} \quad x \in \tilde{\mathcal{D}} \cap \mathcal{E}^+.
\]

Then the closure of \( (A + B, \mathcal{D}) \) in \( \mathcal{E} \) is the generator of a stochastic semigroup, \( S_\mathcal{E} = \{ S_\mathcal{E}(t) \}_{t \geq 0} \), on \( \mathcal{E} \) which leaves \( \tilde{\mathcal{E}} \) invariant. The restrictions \( S_\mathcal{E}(t) := S_\mathcal{E}(t)|_{\tilde{\mathcal{E}}} \), \( t \geq 0 \) constitute a \( C_0 \)-semigroup, \( S_\mathcal{E} \), on \( \tilde{\mathcal{E}} \) generated by the trace of the generator of \( S_\mathcal{E} \) in \( \tilde{\mathcal{E}} \).

**Remark 7.3.** Without assuming item (iv) above one can only guarantee that an extension of \( (A + B, \mathcal{D}) \) is the generator of a substochastic semigroup on \( \mathcal{E} \), which corresponds to a dishonesty of the evolution described by this semigroup. More on this item can be found in [2].

Now we turn to constructing the semigroups \( S^\alpha \).

### 7.1.2 The Banach spaces of measures

Let \( \mathcal{M} \) be the linear space of finite signed measures on \( (\Gamma, B(\Gamma)) \), see [12, Chapter 4]. That is, \( \mu \in \mathcal{M} \) is a \( \sigma \)-additive map \( \mu : B(\Gamma) \to \mathbb{R} \) which takes only finite values. By \( \mathcal{M}^+ \) we denote the set of all such \( \mu \) that take only nonnegative values. Then the Jordan decomposition of \( \mu \) is the unique representation \( \mu = \mu^+ - \mu^- \) with \( \mu^\pm \in \mathcal{M}^+ \). Thus, \( \mathcal{M}^+ \) is a generating cone. Set \( |\mu| = \mu^+ + \mu^- \). Then

\[
\| \mu \| := |\mu|(\Gamma) \quad \text{(7.3)}
\]
A Markov process for a particle system

is a norm, that is clearly additive on $\mathcal{M}^+$. By [12, Proposition 4.1.8, page 119] with this norm $\mathcal{M}$ is a Banach space. Let $\Psi_1$ be the function defined in (2.17). For $n \in \mathbb{N}$, let $\mathcal{M}_n$ be the subset of $\mathcal{M}$ consisting of all those $\mu$ for which $\Psi_1^\mu$ are finite signed measures. Recall that $\Psi_1 = 1 + \Psi$, see (2.14). We equip $\mathcal{M}_n$ with the norm

$$\| \mu \|_n = \int_{\Gamma} \Psi_1^\mu(\gamma) |\mu|(d\gamma) =: \varphi_n(\|\mu\|). \quad (7.4)$$

By the same [12, Proposition 4.1.8, page 119] with this norm $\mathcal{M}_n$ is a Banach space. Now for $\beta > 0$, let $\mathcal{M}_\beta$ be the subset of $\mathcal{M}$ the elements of which remain finite measures being multiplied by $\exp(\beta \Psi(\gamma))$. We equip it with the norm

$$\| \mu \|_\beta = \int_{\Gamma} \exp(\beta \Psi(\gamma)) |\mu|(d\gamma) =: \varphi_\beta(\|\mu\|).$$

Then also $(\mathcal{M}_\beta, \| \cdot \|_\beta)$ is a Banach space. By (2.20) and (2.21) it follows that

$$\forall \mu \in \mathcal{M}_1 \quad |\mu|(\Gamma_\ast) = |\mu|(\Gamma).$$

That is, for each $\mu \in \mathcal{M}_1$, it follows that $|\mu|(\Gamma_\ast) = 0$. Define

$$\mathcal{M}_\ast = \{ \mu \in \mathcal{M} : |\mu|(\Gamma_\ast) = 0 \}.$$ 

Thus, $\mathcal{M}_1 \subset \mathcal{M}_\ast$. Obviously, also all $\mathcal{M}_n$ and $\mathcal{M}_\beta$ have the same property. For a subset, $\mathcal{M}' \subset \mathcal{M}$, let $\mathcal{M}'$ denote its closure in $\| \cdot \|$ defined in (7.3).

**Lemma 7.4.** For each $n \in \mathbb{N}$ and $\beta > 0$, it follows that

$$\overline{\mathcal{M}_n} = \overline{\mathcal{M}_\beta} = \mathcal{M}_\ast. \quad (7.5)$$

**Proof.** Obviously, for each $n \in \mathbb{N}$ and $\beta > 0$, the following holds $\mathcal{M}_\beta \subset \mathcal{M}_n$. Then it is enough to prove the validity of (7.5) for $\mathcal{M}_\beta$. Let us prove the inclusion $\overline{\mathcal{M}_\beta} \subset \mathcal{M}_\ast$. For a given $\mu \in \overline{\mathcal{M}_\beta}$, let $\{ \mu_n \}_{n \in \mathbb{N}} \subset \mathcal{M}_\beta$ be a sequence such that $\| \mu - \mu_n \| \to 0$. Fix $n$ and let then $\Gamma = P \cup N$ be the Hahn decomposition for $\mu - \mu_n$, i.e., $\mu(A) \geq \mu_n(A)$ for each $A \subset P$, and $\mu(A) \leq \mu_n(A)$ for each $A \subset N$. Then

$$\| \mu - \mu_n \| = (\mu - \mu_n)(P) + (\mu - \mu_n)(N) \geq (\mu - \mu_n)(P \cap \Gamma_\ast) + (\mu_n - \mu)(N \cap \Gamma_\ast)$$

$$= (\mu(P \cap \Gamma_\ast) - \mu(N \cap \Gamma_\ast) = \mu^-(\Gamma_\ast) - \mu^+(\Gamma_\ast)) = |\mu|(\Gamma_\ast),$$

where we have taken into account that $|\mu_n|(\Gamma_\ast) = 0$. Then the assumed convergence $\mu_n \to \mu$ yields that $\mu \in \mathcal{M}_\ast$. To prove the opposite inclusion we take an arbitrary $\mu \in \mathcal{M}_\ast$ and write its Jordan decomposition $\mu = \mu^+ - \mu^-$. For a given $n \in \mathbb{N}$, let $I_n$ be the indicator of the set $\Gamma_{\ast,n}$ defined in (2.20). Then both $\mu^\pm := I_n \mu^\pm$ are in $\mathcal{M}_\beta$. At the same time, by (2.20) the sequence of function $J_n(\gamma) := 1 - I_n(\gamma)$ converges to zero pointwise on $\Gamma_{\ast}$. Since $\mu \in \mathcal{M}_\ast$, we have

$$\| \mu^\pm - \mu^\pm_n \| = \int_{\Gamma} J_n(\gamma) \mu^\pm(d\gamma) = \int_{\Gamma_{\ast}} J_n(\gamma) \mu^\pm(d\gamma) \to 0, \quad \text{as} \ n \to +\infty. \quad (7.6)$$

By the triangle inequality we then obtain that $\| \mu - \mu_n \| \to 0$, where $\mu_n := \mu_n^+ - \mu_n^- \in \mathcal{M}_\beta$.

By the very definition of the spaces $\mathcal{M}_n$, $\mathcal{M}_\beta$ and $\mathcal{M}_\ast$, we conclude that they have generating cones of positive elements consisting of those $\mu$ that take nonnegative values only. 

EJP 26 (2021), paper 72. https://www.imstat.org/ejp

Page 39/53
A Markov process for a particle system

**Corollary 7.5.** The set \( M_* \) equipped with the norm \( \| \cdot \| \) defined in (7.3) is a Banach space. Let \( \mathcal{M}_+^\ast \) be its cone of positive elements. Then for each \( n \in \mathbb{N} \) and \( \beta > 0 \), it follows that

\[
\overline{\mathcal{M}_n^\ast} = \overline{\mathcal{M}_\beta^\ast} = \mathcal{M}_+^\ast,
\]

where we mean the closure in the norm of \( \mathcal{M}_* \).

**Proof.** The first part of the statement follows directly by (7.5). The second part is obtained by the construction used in (7.6).

Let \( \mathcal{M}_+^{\ast,1} \) be the subset of \( \mathcal{M}_* \) consisting of probability measures, i.e., for which \( \| \mu \| = \mu(\Gamma) = \mu(\Gamma_*) = 1 \). Then by (2.21) it follows that

\[
\mathcal{P}_{\exp} \subset \mathcal{M}_+^{\ast,1} \subset \mathcal{M}_+^1 = \mathcal{M}_1^\ast, 
\]

By (2.19), for each \( \beta > 0 \), we also have

\[
\mathcal{P}_{\exp} \subset \mathcal{M}_+^{\ast,1} : = \mathcal{M}_\beta \cap \mathcal{M}_+^{\ast,1} \subset \mathcal{M}_n \cap \mathcal{M}_+^{\ast,1}, \quad \text{for all } n \in \mathbb{N}. 
\]

### 7.1.3 The stochastic semigroup

For a given \( \alpha \in (0,1] \), set

\[
\Phi_\alpha(\gamma) = \sum_{x \in \gamma} \int_{\mathbb{R}^d} a_\alpha(x,y) \exp \left( - \sum_{z \in \gamma \setminus x} \phi(y - z) \right) dy, \quad \gamma \in \Gamma_*.
\]

(7.7)

Since \( \psi(x) < \psi_\alpha(x) \leq \psi(x)/\alpha \), see (2.17), for all \( \alpha \in (0,1] \) we have

\[
\Psi(\gamma) < \Phi_\alpha(\gamma) \leq \Psi(\gamma)/\alpha,
\]

and hence \( \Phi_\alpha(\gamma) < \infty \) for \( \gamma \in \Gamma_* \). Now let \( L^\alpha \) be the corresponding Kolmogorov operator (6.3). Our aim is to define its ‘predual’, \( L^{1,\alpha} \), acting according to the rule

\[
\mu(L^\alpha F) = (L^{1,\alpha} \mu)(F),
\]

(7.9)

for appropriate \( \mu \in \mathcal{P}(\Gamma_*) \) and \( F : \Gamma_* \to \mathbb{R} \), and then to use it to define the corresponding operators acting in the spaces of measures just introduced. Obviously, we can restrict ourselves to the elements of \( \mathcal{M}_* \). By (6.3) and (6.2) we thus obtain it in the form

\[
L^{1,\alpha} = A + B
\]

(7.10)

where \( A \) is the multiplication operator by the function \( -\Phi_\alpha \) defined in (7.7). In view of (7.8) the domain of \( A \) is to be

\[
\mathcal{D} = \{ \mu \in \mathcal{M}_* : \Phi_\alpha \mu \in \mathcal{M}_* \} = \mathcal{M}_1.
\]

(7.11)

To define \( B \) we introduce the following measure kernel

\[
\Omega_\alpha(\Lambda) = \sum_{x \in \gamma} \int_{\mathbb{R}^d} a_\alpha(x,y) \exp \left( - \sum_{z \in \gamma \setminus x} \phi(y - z) \right) \mathbf{1}_{\Lambda}(\gamma \setminus x \cup y) dy,
\]

(7.12)

with \( \gamma \in \Gamma_* \) and \( \Lambda \in \mathcal{B}(\Gamma_*) \). By (7.7) we then have

\[
\Omega_\alpha(\Gamma_*) = \Phi_\alpha(\gamma).
\]

(7.13)
We apply this and (7.8) in (7.17) and obtain
\[ (B\lambda)(\lambda) = \int_{\Gamma_*} \Omega_{\alpha}^*(\lambda)\lambda(d\gamma). \] (7.14)

Note that
\[ B : \mathcal{M}_1^+ \to \mathcal{M}_1^+. \] (7.15)

Moreover, for \( \mu \in \mathcal{M}_1^+ \), by (7.13) and (7.14) we have
\[ \|B\mu\| = (B\mu)(\Gamma_*) = \int_{\Gamma_*} \Phi_{\alpha}(\gamma)\mu(d\gamma) = -(A\mu)(\Gamma_*). \] (7.16)

Hence, we can take \( \mathcal{M}_1 \) as the domain of \( B \) and then define \( L^{1,\alpha} \) by (7.10) with domain \( \mathcal{D} = \mathcal{M}_1 \), see (7.11).

In the sequel, we will use one more property of \( B \). By (7.12), (7.14) and (7.4) we get
\[ \varphi_n(B\mu) = \int_{\Gamma_*} \Psi_1(\gamma)(B\mu)(d\gamma) \] (7.17)
\[ = \int_{\Gamma_*} \left( \sum_{x \in \gamma} \int_{B^1} a_n(x, y) \exp \left( -\sum_{z \in \gamma \setminus x} \phi(y-z) \right) \Psi_1(\gamma \setminus x \cup y)dy \right) \mu(d\gamma). \]

By (2.17) it follows that
\[ \Psi_1(\gamma \setminus x \cup y) = (\Psi_1(\gamma) + \psi(y) - \psi(x))^n \leq 2^n \Psi_1(\gamma). \] (7.18)

We apply this and (7.8) in (7.17) and obtain
\[ \forall \mu \in \mathcal{M}_{n+1}^+ \quad \|B\mu\| = \varphi_n(B\mu) \leq \left( \frac{2^n}{\alpha} \right) \|\mu\|_{n+1}. \]

This yields the following extension of (7.15)
\[ B : \mathcal{M}_{n+1}^+ \to \mathcal{M}_n^+, \] (7.19)

holding for all \( n \in \mathbb{N} \). Since \( \|A\mu\|_n \leq \alpha^{-1}\|\mu\|_{n+1} \), by (7.19) we also get
\[ \forall n \in \mathbb{N}_0 \quad L^{1,\alpha} : \mathcal{M}_{n+1} \to \mathcal{M}_n, \]

that can be used to define the powers of \( L^{1,\alpha} \)
\[ (L^{1,\alpha})^n : \mathcal{M}_{n+m} \to \mathcal{M}_n, \quad n \in \mathbb{N}_0, \quad m \in \mathbb{N}. \] (7.20)

Here – and in the sequel in similar expressions – \( \mathcal{M}_0 = \{0\} \) (corresponding to \( \mathcal{M}_n \) with \( n = 0 \)) is understood as \( \mathcal{M}_1 \). Let us now define a bounded linear operator \( L^{1,\alpha}_{\beta,\beta'} : \mathcal{M}_\beta \to \mathcal{M}_{\beta'}, \beta' < \beta, \) the action of which is the same as that of the unbounded operator \( L^{1,\alpha}_\beta = A + B \) defined in (7.10) and (7.14). For a given \( \mu \in \mathcal{M}_\beta \), let \( \mu = \mu^+ - \mu^- \) be its Jordan decomposition. Then
\[ L^{1,\alpha}_\mu = (B\mu^+ - A\mu^-) - (B\mu^- - A\mu^+) =: \mu^+ - \mu^- \in \mathcal{M}_\beta^+. \]

This yields that
\[ \|L^{1,\alpha}_\mu\|_{\beta'} \leq \|\mu^+\|_{\beta'} + \|\mu^-\|_{\beta'} = \|B\mu^+\|_{\beta'} + \|B\mu^-\|_{\beta'} + \|A\mu^+\|_{\beta'} + \|A\mu^-\|_{\beta'}, \] (7.21)

holding for all \( \mu \in \mathcal{M}_\beta \). Here we have used the additivity of the norms on the positive cone as well as the positivity of \( B \) and \( -A \). By (7.8) and the following evident inequality
\[ xe^{-\alpha x} \leq 1/e \kappa \] holding for all positive \( x \) and \( \kappa \), we obtain
\[ \Phi_{\alpha}(\gamma) \exp (\beta' \Psi_0(\gamma)) \leq \frac{1}{\alpha e(\beta - \beta')} \exp (\beta \Psi(\gamma)). \] (7.22)
A Markov process for a particle system

By (7.22) for \( \mu \in \mathcal{M}_\beta^+ \), we then get

\[
\|A\mu\|_{\beta'} \leq \frac{\|\mu\|_{\beta}}{\alpha e (\beta - \beta')}.
\] (7.23)

Next, similarly as in (7.17) it follows that

\[
\int_{\Gamma_*} \exp (\beta' \Psi(\gamma)) (B\mu)(d\gamma)
= \int_{\Gamma_*} \left( \sum_{x \in \gamma} \int_{\mathbb{R}^d} a_\alpha(x, y) \exp \left( - \sum_{z \in \gamma \setminus x} \phi(y - z) \right) \right.
\]
\[
\times \exp (\beta' \Psi(\gamma)) \exp (\beta' [\psi(y) - \psi(x)]) dy \mu(d\gamma)
\leq e^\beta \int_{\Gamma_*} \Phi_\alpha(\gamma) \exp (\beta' \Psi(\gamma)) \mu(d\gamma) \leq \frac{e^\beta \|\mu\|_{\beta}}{\alpha e (\beta - \beta')}.
\]

We combine this estimate with (7.23) and (7.21) to obtain

\[
\|L_{\beta, \alpha}^{+} \| \leq \frac{e^\beta + 1}{\alpha e (\beta - \beta')},
\]

In a similar way, for each \( n \in \mathbb{N} \), we also obtain, cf. (4.13),

\[
\|L_{\beta, \alpha}^{+} \|^n \leq \left( \frac{n}{e T_\alpha(\beta, \beta')} \right)^n, \quad T_\alpha(\beta, \beta') := \frac{\alpha (\beta - \beta')}{e^\beta + 1}.
\] (7.24)

By (7.20), for each \( n \in \mathbb{N} \) and \( \mu \in \mathcal{M}_\beta \), we have that \( (L_{\beta, \alpha}^{+})^n \mu \in \mathcal{M}_{\beta'} \), \( \beta' < \beta \), and the following holds

\[
(L_{\beta, \alpha}^{+})^n \mu = (L_{\beta, \alpha}^{+})^n \mu.
\] (7.25)

**Lemma 7.6.** For each \( \alpha \in (0, 1] \), the closure of \( (L_{\beta, \alpha}^{+}, \mathcal{M}_\beta) \) in \( \mathcal{M}_\star \) is the generator of a stochastic semigroup, \( S^\alpha = \{S^\alpha(t)\}_{t \geq 0}, \) in \( \mathcal{M}_\beta \) such that \( S^\alpha(t) : \mathcal{M}_n \to \mathcal{M}_n \) for each \( n \in \mathbb{N} \). The restrictions \( S^\alpha(t)|_{\mathcal{M}_n} \) constitute a \( C_0 \)-semigroup on \( \mathcal{M}_n \). Moreover, for each \( \beta > 0 \) and \( \beta' \in (0, \beta) \), \( S^\alpha(t) : \mathcal{M}_\beta^{+} \to \mathcal{M}_\beta^{+} \) for \( t < T_\alpha(\beta, \beta') \), see (7.24).

**Proof.** The construction of the semigroup in question will be made, in particular, by showing that all the conditions of Proposition 7.2 are met. We thus begin by checking whether each of the spaces \( \mathcal{M}_n \) and \( \mathcal{M}_\beta \) enjoys the properties listed in Assumption 7.1. By Lemma 7.4 the density assumed in (i) is guaranteed. Each of these spaces is a Banach space with the corresponding norm, that was already mentioned in the course of their introduction. The properties assumed in (iii) are evident, whereas (iv) follows by Corollary 7.5. Thus, we can start checking the validity of the conditions imposed in Proposition 7.2. Recall that both \( A \) and \( B \) are (densely) defined on the domain \( D = \mathcal{M}_1 \), see (7.11) and Lemma 7.4, and \( A \) is the multiplication operator by the function \( (-\Phi_\alpha) \). Hence, condition (i) of Proposition 7.2 is satisfied. Moreover, \( A \) generates the semigroup \( S \) consisting of the following operators

\[
(S(t)\mu)(d\gamma) = \exp (-t\Phi_\alpha(\gamma)) \mu(d\gamma).
\] (7.26)

Then

\[
\|S(t)\mu\| \leq \|\mu\|,
\] (7.27)
A Markov process for a particle system

which obviously holds for all \( \mu \in \mathcal{M}_* \). To check whether \( S \) is strongly continuous in \( \mathcal{M}_* \), for a given \( \mu \in \mathcal{M}_* \) and \( \epsilon > 0 \), we have to find \( \delta > 0 \) such that \( \| \mu - S(t)\mu \| < \epsilon \) for all \( t < \delta \). Since \( \mathcal{M}_* \) is the \( \| \cdot \| \)-closure of \( \mathcal{M}_1 \) (by Lemma 7.4), for the chosen \( \mu \), one finds \( \mu' \in \mathcal{M}_1 \) such that \( \| \mu - \mu' \| < \epsilon/3 \). Then by (7.26) and (7.27) we get

\[
\| \mu - S(t)\mu \| \leq \| \mu - \mu' \| + \| S(t)(\mu - \mu') \| + \| \mu' - S(t)\mu' \| \leq t\| A\mu' \| + 2\epsilon/3 \leq (t/\alpha)\| \mu' \| + 2\epsilon/3,
\]

which completes the proof for \( \mathcal{M}_* \). Clearly, \( S(t) : \mathcal{M}_n^+ \to \mathcal{M}_n^{+\ast} \), and the domain of the trace of \( A \) in \( \mathcal{M}_n \) is \( \mathcal{D}_n = \mathcal{M}_{n+1} \). Then the proof that \( S(t)|_{\mathcal{M}_n} \) is strongly continuous in \( \mathcal{M}_n \) can be performed similarly as in (7.28). Thus, condition (ii) of Proposition 7.2 is met. In view of (7.19) to complete the proof of (iii) we have to show \( \varphi((A + B)\mu) = 0 \) whenever \( \mu \in \mathcal{M}_n^{+\ast} \), which is obviously the case by (7.16). Then it remains to show that, for a fixed \( n \in \mathbb{N} \),

\[
\int_{\Gamma_*} \Psi_1^n(\gamma)(L^{1,\alpha}\mu)(d\gamma) \leq c \int_{\Gamma_*} \Psi_1^n(\gamma)\mu(d\gamma) - \epsilon \int_{\Gamma_*} \Phi_\alpha(\gamma)\mu(d\gamma),
\]

holding for each \( \mu \in \mathcal{M}_n^{+\ast} \) and some positive \( c \) and \( \epsilon \), possibly dependent on \( n \). In view of the following estimate, cf. (7.8),

\[
\alpha\Phi_\alpha(\gamma) \leq 1 + n \sum_{x \in \gamma} |\psi(x)| \leq \Psi_1^n(\gamma), \quad n \in \mathbb{N}, \quad \gamma \in \Gamma_*,
\]

it is enough to show (7.29) with \( \epsilon = 0 \) and sufficiently big \( c \). By (7.9) this amounts to showing

\[
L^n\Psi_1^n(\gamma) \leq c\Psi_1^n(\gamma), \quad \gamma \in \Gamma_*.
\]

By (7.18) it follows that

\[
|\Psi_1^n(\gamma \setminus x \cup y) - \Psi_1^n(\gamma)| \leq 2^n|\psi(y) - \psi(x)|\Psi_1^{n-1}(\gamma).
\]

Assume that \( |x| > |y| \). By (2.17) we have

\[
|\psi(y) - \psi(x)| = (|x|^{d+1} - |y|^{d+1}) \psi(x)\psi(y)
\]

\[
\leq \left[ (|x - y| + |y|)^{d+1} - |y|^{d+1} \right] \psi(x)\psi(y)
\]

\[
= \sum_{l=1}^{d+1} \binom{d+1}{l} |x - y|^l |y|^{d+1-l} \psi(x)\psi(y)
\]

\[
\leq \psi(x) \sum_{l=1}^{d+1} \binom{d+1}{l} |x - y|^l = \psi(x)v(|x - y|).
\]

For \( |y| > |x| \), in a similar way we get

\[
|\psi(y) - \psi(x)| \leq \sum_{l=1}^{d+1} \binom{d+1}{l} |x - y|^l |x|^{d+1-l} \psi(x)\psi(y)
\]

\[
\leq \sum_{l=1}^{d+1} \binom{d+1}{l} |x - y|^l |y|^{d+1-l} \psi(x)\psi(y)
\]

\[
\leq \psi(x) \sum_{l=1}^{d+1} \binom{d+1}{l} |x - y|^l = \psi(x)v(|x - y|).
\]
Now we apply (7.31), (7.32) and (7.33) to obtain
\[
\text{LHS}(7.30) \leq 2^n \sum_{x \in \gamma} \int_{\mathbb{R}^d} \psi_\alpha(x) \psi(x) a(x-y) v(|x-y|) \Psi_1^{n-1}(\gamma) dy
\]
\[
\leq 2^n \Psi_1^{n-1}(\gamma) \left( \sum_{x \in \gamma} \psi(x) \right) \int_{\mathbb{R}^d} a(y) v(|y|) dy
\]
\[
\leq \left( 2^n \int_{\mathbb{R}^d} a(y) v(|y|) dy \right) \Psi_n(\gamma),
\]
that by (3.3) proves (7.30). Thus, all the conditions of Proposition 7.2 are met, which proves the part of the lemma related to the stochastic semigroup $S^\alpha$ acting in $\mathcal{M}_\alpha$ and its restrictions to $\mathcal{M}_n, n \in \mathbb{N}$. To prove the second part of the lemma we use the estimate in (7.24) and define bounded operators $S^\alpha_{\beta, \beta'}(t) : \mathcal{M}_\beta \to \mathcal{M}_{\beta'}, t < T_\alpha(\beta, \beta')$ by the series
\[
S^\alpha_{\beta, \beta'}(t) = \mathbb{I}_{\beta, \beta} + \sum_{n=1}^{\infty} t^n (L_{\beta, \beta'}^1)^n,
\]
where $\mathbb{I}_{\beta, \beta}$ is the embedding operator. By the latter formula and (7.25) we conclude that
\[
\forall \mu \in \mathcal{M}_\beta \quad S^\alpha(t) \mu = S^\alpha_{\beta, \beta'}(t) \mu, \quad t < T_\alpha(\beta, \beta'). \tag{7.34}
\]
By (7.24) $S^\alpha_{\beta, \beta'}(t) : \mathcal{M}_\beta \to \mathcal{M}_{\beta'}$ is a bounded operator, the norm of which satisfies
\[
\|S^\alpha_{\beta, \beta'}(t)\| \leq \frac{T_\alpha(\beta, \beta')}{T_\alpha(\beta, \beta') - t}.
\]
The positivity of $S^\alpha_{\beta, \beta'}(t)$ follows by (7.34) and the positivity of $S^\alpha(t)$. This completes the proof.

In view of (7.1), we conclude that Lemma 7.6 establishes the existence of the transition function corresponding to $L^\alpha$, with the properties arising from the corresponding properties of the semigroup. Note that $\delta_\gamma \in \mathcal{M}_\alpha$ (and hence in all $\mathcal{M}_n$ and $\mathcal{M}_\beta$) if and only if $\gamma \in \Gamma_n$, that will be assumed below. It is straightforward that $p^\alpha_\gamma(\cdot, \cdot)$ satisfies the corresponding standard conditions and thus determines finite-dimensional distributions of a Markov process, see [15, pages 156, 157]. Our next step is to show that it has cadlag versions.

### 7.2 Constructing path measures

The construction of the families of path measures $P^\alpha$ which solve the martingale problem in the sense of Definition 3.3 can be done by defining their finite dimensional marginals with the help of the transition function (7.1). In this case, however, the one dimensional marginals
\[
\Pi^\alpha = S^\alpha(t) \mu = \int_{\Gamma_n} p^\alpha_{\gamma}(\cdot, \cdot) \mu(\gamma), \tag{7.35}
\]
need not be in $\mathcal{P}_{\exp}$, even for $\mu \in \mathcal{P}_{\exp}$. The only fact guaranteed by Lemma 7.6 is that $\Pi^\alpha_n \in \mathcal{M}_n$, for all $n \in \mathbb{N}$, and that $\Pi^\alpha_t \in \mathcal{M}_\beta$ with $t$ belonging to a bounded interval. This obstacle is removed by the following statement.

**Lemma 7.7.** For a given $\mu \in \mathcal{P}_{\exp}$, let $\{\mu^\alpha_t : \mu_0 = \mu, \ t \geq 0\} \subset \mathcal{P}_{\exp}$ be the family of measures defined by their correlation functions $k^\alpha_t$ according to (6.5). For the same $\mu$, let $\Pi^\alpha_t, \ t \geq 0$ be as in (7.35). Then $\Pi^\alpha_t = \mu^\alpha_t$ for all $t > 0$ and $\alpha \in (0, 1).$
A Markov process for a particle system

**Proof.** Exactly as in Theorem 5.2 one proves that the Fokker-Planck equation (3.18) with $L$ replaced by $L^\alpha$ has a unique solution, which is $\mu^\alpha_t$. At the same time, by construction $\Pi^\alpha_t$ also solves this equation.

Now we can start constructing the path measures in question. To this end we use Chentsov’s theorem in the following version, see [15, Theorems 8.6 – 8.8, pages 137–139]. Recall that the complete metric $\upsilon_\ast$ of $\Gamma_\ast$ was introduced in (2.24). For $\alpha \in (0, 1]$, $\gamma \in \Gamma_\ast$ and $u, v \geq 0$, set

$$w^\alpha_u(\gamma) = \int_{\Gamma_\ast} v_\ast(\gamma, \gamma') p^\alpha_u(\gamma, d\gamma').$$  \hspace{1cm} (7.36)

$$W^\alpha_{u, v}(\gamma) = \int_{\Gamma_\ast} v_\ast(\gamma, \gamma') w^\alpha_u(\gamma') p^\alpha_v(\gamma', d\gamma').$$

Thereafter, for $t_3 \geq t_2 \geq t_1 \geq 0$ (such sets are called **triples**), let us consider

$$\mathcal{W}^\alpha(t_1, t_2, t_3) = \int_{\Gamma_\ast} W^\alpha_{t_3-t_2, t_2-t_1}(\gamma) \Pi^\alpha_{t_1}(d\gamma') = \int_{\Gamma_\ast} W^\alpha_{t_3-t_2, t_2-t_1}(\gamma) \mu^\alpha_{t_1}(d\gamma'),$$  \hspace{1cm} (7.37)

where $\mu$, $\mu^\alpha$ and $\Pi^\alpha$ are as in (7.35).

**Proposition 7.8.** (Chentsov) Assume that there exists $C_\alpha > 0$ and $\delta > 0$ such that, for each triple $t_1, t_2, t_3$, the following holds

$$\mathcal{W}^\alpha(t_1, t_2, t_3) \leq C_\alpha |t_3 - t_1|^2, \quad t_3 - t_1 < \delta.$$  \hspace{1cm} (7.38)

Then the following is true:

(i) The transition function (7.1) and $\mu \in \mathcal{P}_{\exp}$ determine a probability measure $P^\alpha$ on $\mathcal{D}_{\mathbb{R}_+}(\Gamma_\ast)$.

(ii) If the estimate in (7.38) holds uniformly in $\alpha$, i.e., with some $C > 0$ independent of $\alpha \in (0, 1]$, and if the family $\{\Pi^\alpha : \alpha \in (0, 1]\} \subset \mathcal{P}(\Gamma_\ast)$ is tight for each $t > 0$, then the family $\{P^\alpha : \alpha \in (0, 1]\}$ of measures as in (i) is also tight, and hence possesses accumulation point in the weak topology.

Note that the tightness of the family $\{\Pi^\alpha : \alpha \in (0, 1]\}$ follows by Lemmas 7.7 and 6.2.

**Lemma 7.9.** For each $\mu \in \mathcal{P}_{\exp}$, the estimate in (7.38) holds true for all $\alpha \in (0, 1]$ with one and the same $C > 0$.

**Proof.** By (7.1) and standard semigroup formulas, e.g., [15, page 9], we have

$$p^\alpha_u(\gamma, \cdot) = \delta_\gamma + \int_0^u L^{1-\alpha} p^\alpha_u(\gamma, \cdot) ds,$$  \hspace{1cm} (7.39)

since $\delta_\gamma \in \mathcal{D} = M_1$. Then by this formula and (7.36) we obtain

$$w^\alpha_u(\gamma) = w^0_u(\gamma) + \int_0^u \left( \int_{\Gamma_\ast} v_\ast(\gamma, \gamma')(L^{1-\alpha} p^\alpha_u(\gamma, d\gamma')) ds \right)$$

$$= \int_0^u \left( \int_{\Gamma_\ast} (L^\alpha v_\ast(\gamma, \gamma') p^\alpha_u(\gamma, d\gamma')) ds, \right.$$  \hspace{1cm} (7.40)

where we have taken into account that $w^0_u(\gamma) = v_\ast(\gamma, \gamma) = 0$ as $v_\ast$ is a metric. We apply now $L^\alpha$ to $v_\ast(\gamma, \cdot)$ – which is a bounded continuous function of $\gamma'$, and obtain

$$J^\gamma(\cdot) := (L^\alpha v_\ast)(\gamma, \gamma')$$

$$= \sum_{x \in Y} \int_{\mathbb{R}^d} a(x, y) \exp \left( - \sum_{z \in \gamma' \setminus x} \phi(y - z) \right) [v_\ast(\gamma, \gamma' \setminus x \cup y) - v_\ast(\gamma, \gamma')] dy.$$
A Markov process for a particle system

By the triangle inequality for \( v \), we then get from the latter

\[
|J^i(\gamma')| \leq \sum_{x \in \gamma'} \int_{\mathbb{R}^d} a_u(x, y) v_u(\gamma' \setminus x \cup y, \gamma')dy. \tag{7.41}
\]

In view of (2.24), to estimate \( v_u(\gamma', x \cup y, \gamma') \) we consider \( |\theta(y) - \theta(x)| \) with \( \theta(x) = g(x)\psi(x) \), \( g \in C^\infty_{\text{L}}(\mathbb{R}^d) \), \( \|g\|_{B.L} \leq 1 \), for which we obtain, cf. (3.7),

\[
|\theta(y) - \theta(x)| = \psi(x)\psi(y) \left| \frac{g(y) - g(x)}{\psi(x)} - \frac{g(x)}{\psi(y)} \right| \tag{7.42}
\]

\[
= \psi(x)\psi(y) \left| \frac{g(y) - g(x)}{\psi(y)} + g(y) \left[ \frac{1}{\psi(x)} - \frac{1}{\psi(y)} \right] \right|
\]

\[
\leq \psi(x)|x - y| + \psi(x)\psi(y)|x|^{d+1} - |x|^{d+1} \]

\[
\leq \psi(x) \left[ |x - y| + \sum_{l=1}^{d+1} \left( \frac{d+1}{l} \right) |x - y|^l \right].
\]

Now we use this in (7.41) and arrive at

\[
|J^i(\gamma')| \leq C_a \Psi(\gamma'), \quad C_a := m_1^a + \sum_{l=1}^{d+1} \left( \frac{d+1}{l} \right) m_1^a. \tag{7.43}
\]

Then we use (7.43) in (7.40) and obtain

\[
w_u^a(\gamma) \leq C_a \int_0^u \chi_u^a(\gamma)ds, \quad \chi_u^a(\gamma) := \int_{\Gamma_u} \Psi(\gamma') p_u^a(\gamma, d\gamma'). \tag{7.44}
\]

Note that \( \chi_u^0(\gamma) = \Psi(\gamma) \). Similarly as in (7.39) we write

\[
\chi_u^a(\gamma) = \Psi(\gamma) + \int_0^u \left( \int_{\Gamma_u} (L^a \Psi)(\gamma') p_u^a(\gamma, d\gamma') \right) dv. \tag{7.45}
\]

Like in (7.30) one gets

\[
(L^a \Psi)(\gamma) \leq |(L^a \Psi)(\gamma)| \leq 2c_a \Psi(\gamma),
\]

where \( c_a \) is as in (3.9). We use this in (7.45), take also into account the definition of \( \chi_u^a \) in (7.44) and obtain

\[
\chi_u^a(\gamma) \leq \Psi(\gamma) + 2c_a \int_0^u \chi_u^a(\gamma)dv,
\]

which by the Grönwall inequality and (7.44) yields the following estimate

\[
w_u^a(\gamma) \leq C_a e^{2c_a u} \Psi(\gamma).
\]

We employ this in the second line of (7.36) and obtain

\[
W_{w,u}^a(\gamma) \leq C_a e^{2c_a u} q_u^a(\gamma), \quad q_u^a(\gamma) := \int_{\Gamma_u} \Psi(\gamma') v_u(\gamma, \gamma') p_u^a(\gamma, d\gamma'). \tag{7.46}
\]

Note that \( q_u^0(\gamma) = 0 \) as \( v_u \) is a metric. Similarly as in (7.40) we then get

\[
q_u^a(\gamma) = \int_0^u \left( \int_{\Gamma_u} (L^a \Psi v_u(\gamma, \cdot))(\gamma') p_u^a(\gamma, d\gamma') \right) dv.
\]
Thus, we have to estimate

\[
| (L^\alpha \Psi_{\nu_\ast}(\gamma, \cdot))(\gamma') | \leq \sum_{x \in \gamma'} \int_{\mathbb{R}^d} a(x - y) \left| \Psi(\gamma' \setminus x \cup y) \nu_\ast(\gamma, \gamma' \setminus x \cup y) \right| dy
\]

(7.47)

\[
- \Psi(\gamma') \nu_\ast(\gamma, \gamma') \left[ dy \right] \leq \sum_{x \in \gamma'} \int_{\mathbb{R}^d} a(x - y)|\psi(y) - \psi(x)| dy
\]

\[
+ \Psi(\gamma') \sum_{x \in \gamma'} \int_{\mathbb{R}^d} a(x - y) \nu_\ast(\gamma, \gamma' \setminus x \cup y) dy
\]

\[
\leq c_a \Psi(\gamma') + c_a \Psi^2(\gamma'),
\]

where we used the same estimate as in as in (7.42). Now we use (7.47) in (7.46) and then plug this into (7.37). In doing so, we will deal with

\[
\int_{\Gamma_+} p^\alpha(\gamma, d\gamma') \Pi^\alpha_{t_1}(d\gamma) = \mu^\alpha_{t_1+s}(d\gamma'),
\]

that follows by the Chapman-Kolmogorov property of the transition function (7.1), see [15, page 156], and then by Lemma 7.7. Thereafter, we obtain

\[
\mathcal{W}(t_1, t_1 + v, t_1 + u + v) \leq C_a u e^{\frac{2}{a} u} \int_0^v \left( c_a \mu^\alpha_{t_1+s}(\Psi) + c_a \mu^\alpha_{t_1+s}(\Psi^2) \right) ds.
\]

(7.48)

We recall that \( \mu \in \mathcal{P}^0_{\text{exp}} \) and thus the correlation functions of \( \mu^\alpha \) satisfy the estimate in claim (a) of Proposition 4.2 with \( \vartheta = \vartheta_0 + t \). Then by (2.18) we get

\[
\mu^\alpha(\Psi) \leq \langle \psi \rangle e^{\vartheta_0 + t}, \quad \mu^\alpha(\Psi^2) \leq \langle \psi \rangle e^{\vartheta_0 + t} + \langle \psi \rangle^2 e^{2 \vartheta_0 + 2t}.
\]

We use this in (7.48) and obtain

\[
\mathcal{W}(t_1, t_1 + v, t_1 + u + v) \leq C(u + v)^2,
\]

where, for a fixed \( \delta > 0 \), the independent of \( \alpha \) constant can be calculated explicitly for \( u + v < \delta \). This yields (7.38) with \( C \) independent of \( \alpha \), and hence completes the whole proof.

7.3 Proof of Theorem 6.1

For each \( \alpha \in (0, 1] \), \( s \geq 0 \) and \( \mu \in \mathcal{P}^0_{\text{exp}} \), by Proposition 7.8 the measure \( P^\alpha_{s, \mu} \) on \( \mathcal{D}_{(s, +\infty)}(\Gamma_+) \) is defined by its finite dimensional marginals constructed with the use of the transition function (7.1). Namely, for \( s \leq t_1 < t_2 < \cdots < t_m \) and \( A_1, \ldots, A_m \in B(\Gamma_+) \), we have, cf. [15, eq. (1.10), page 157],

\[
P^\alpha_{s, \mu} \left( (I_{A_1} \circ \omega_{t_1}) \cdots (I_{A_m} \circ \omega_{t_m}) \right) = \int_{\Gamma_+^{m+1}} I_{A_m}(\gamma_m) p^\alpha_{t_m-t_{m-1}}(\gamma_{m-1}, d\gamma_m) \]

(7.49)

\[
\times I_{A_{m-1}}(\gamma_{m-1}) p^\alpha_{t_{m-1}-t_{m-2}}(\gamma_{m-2}, d\gamma_{m-1}) \cdots I_{A_1}(\gamma_1) p^\alpha_{t_1-s}(\gamma_0, d\gamma_1) \mu(d\gamma_0).
\]

In particular, for \( t \geq s \), this yields

\[
P^\alpha_{s, \mu} \circ \omega^{-1}_t = S^\alpha(t-s)\mu.
\]

(7.50)

Then the validity of conditions (a) and (b) of Definition 3.3 follow by (7.50) and Lemma 7.7. Now we turn to proving the validity of (c). Let \( G \) be as in (3.17) with a given
A Markov process for a particle system

$m \in \mathbb{N}$ and $s \leq s_1 < s_2 < \cdots < s_m < t_2$. For a given $F \in \mathcal{D}(L)$ and $u \in [s_m, t_2]$, we set $F_u = F \circ \varpi_u$, $K_u = (LF) \circ \varpi_u$ and $K_u^n = (L^\alpha F) \circ \varpi_u$. Next, we define

$$\chi^\alpha_{s_1}(d\gamma) = C_1 F_1(\gamma) \mu^u_{s_1}(d\gamma) = C_1 F_1(\gamma) \int_{\Gamma} P^\alpha_{s_1-s}(\gamma_0, d\gamma_1) \mu(d\gamma_0), \quad C_1^{-1} := \int_{\Gamma} F_1(\gamma) \mu^u_{s_1}(d\gamma).$$

By (7.35) and Lemma 7.7, and then by claim (iv) of Proposition 3.2, we have that $\chi^\alpha_{s_1} \in \mathcal{P}^{\beta \mu}_{\exp}$ with $\vartheta_l$ dependent on $s_1 - s$ and the type of $\mu \in \mathcal{P}_{\exp}$, and independent of $\alpha$ since the norms of $L^\Delta \alpha$ can be estimated uniformly in $\alpha \in [0, 1]$. Then we define recursively

$$\chi^\alpha_{s_l}(d\gamma) = C_l F_l(\gamma) \int_{\Gamma} P^\alpha_{s_l-s_{l-1}}(\gamma_0, d\gamma_1) \chi^\alpha_{s_{l-1}}(d\gamma_0), \quad l = 2, \ldots, m,$$

and obtain $\chi^\alpha_{s_m} \in \mathcal{P}^{\beta \mu}_{\exp}$ with $\vartheta_m$ independent of $\alpha$. Thereafter, by (7.49) we conclude that

$$P^{\alpha}_{s,\mu}(F_u G) = CP^{\alpha}_{s_m,\chi_m}(F_u) = CP^{\alpha}_{s_m,\chi_m}(F \circ \varpi_u), \quad u \geq s_m, \quad (7.51)$$

where $C$ is a normalizing constant, i.e. $C = P^{\alpha}_{s,\mu}(G)$. Then $P^{\alpha}_{s,\mu}(H) = 0$ follows by the fact that the map $u \mapsto P^{\alpha}_{s,\chi}(\cdot \circ \varpi_{u}^{-1})$ solves the Fokker-Planck equation (3.18) with $L^\alpha$, see (7.50). This proves (c), and hence the family $\{P^{\alpha}_{s,\mu} : s \geq 0, \mu \in \mathcal{P}_{\exp}\}$ is a unique solution of the corresponding restricted martingale problem, see Theorem 5.5.

By Lemma 7.9 and claim (ii) of Proposition 7.8, for each $s$ and $\mu$, the family $\{P^{\alpha}_{s,\mu} : \alpha \in (0, 1]\}$ is relatively weakly compact, and each of its accumulation points has the same one dimensional marginals, that coincide with the measures $\mu_\nu$, see Lemmas 6.2 and 7.7. Let us show that these accumulation points solve the restricted initial value martingale problem for $L$. By Lemmas 6.2 and 7.7 one concludes that conditions (a) and (b) of Definition 3.3 are met, and we thus turn to proving (3.16). Given sequence $\{\alpha_n\}_{n \in \mathbb{N}} \subset (0, 1], \alpha_n \to 0$ and $s \geq 0, \mu \in \mathcal{P}_{\exp}$, let $P^{\alpha_n}_{s,\mu} \Rightarrow P_{s,\mu}$. Let also $G$ in (3.15) be as in (3.17) with a given $m, s_1, \ldots, s_m$ and $F_j \in \mathcal{F}$, $j = 1, \ldots, m$. Set $C_n = P^{\alpha_n}_{s,\mu}(G)$. Then the measures $\nu_{n,u} \in \mathcal{P}(\Gamma_*)$ defined by

$$\nu_{n,u}(\mathcal{A}) = C_n^{-1} P^{\alpha_n}_{s,\mu}(G \cdot (\mathcal{I}_\mathcal{A} \circ \varpi_u)) = P^{\alpha_n}_{s_m,\chi_m}(\varpi_u^{-1}(\mathcal{A})), \quad u \in [s_m, t_2], \quad \mathcal{A} \in \mathcal{B}(\Gamma_*),$$

are in $\mathcal{P}_{\exp}$ with independent of $n$ and $u \in [s_m, t_2]$ (see (7.51)). We also let

$$\nu_u(\mathcal{A}) = C_1^{-1} P^{\alpha}_{s,\mu}(G \cdot (\mathcal{I}_\mathcal{A} \circ \varpi_u)), \quad u \in [s_m, t_2], \quad \mathcal{A} \in \mathcal{B}(\Gamma_*),$$

with $C = P^{\alpha}_{s,\mu}(G)$. Then $\nu_{n,u} \Rightarrow \nu_u$ for all $u \in [s_m, t_2]$. By Lemma 6.4 this yields $\nu_u \in \mathcal{P}_{\exp}$, and hence the corresponding correlation functions satisfy $k^{\alpha_n}_n, k_n \in K_{\theta}$ for all $u \in [s_m, t_2]$ and $n \in \mathbb{N}$. To prove $P_{s,\mu}(H) = 0$ we rewrite it, cf. (3.15),

$$P_{s,\mu}(F_{t_2} G) - P_{s,\mu}(F_{t_1} G) = \int_{t_1}^{t_2} P_{s,\mu}(K_u G) du = 0. \quad (7.52)$$

For $u \in [s_m, t_2]$ and $n \in \mathbb{N}$, we then set

$$a_n(u) = P^{\alpha}_{s,\mu}(F_u G) - P^{\alpha_n}_{s,\mu}(F_u G),$$

$$b_n(u) = P_{s,\mu}(K_u G) - P^{\alpha_n}_{s,\mu}(K_u G),$$

$$c_n(u) = P^{\alpha_n}_{s,\mu}((K_u - K^{\alpha}_n) G).$$

Since $P^{\alpha_n}_{s,\mu}(H) = 0$, it follows that

$$\text{LHS}(7.52) = [a_n(t_2) - a_n(t_1)] = \int_{t_1}^{t_2} b_n(u) du - \int_{t_1}^{t_2} c_n(u) du =: I_n^{(1)} + I_n^{(2)} + I_n^{(3)}. \quad (7.53)$$
A Markov process for a particle system

By the assumed weak convergence of $P^{n_{i+1}}_{s_i}$ one readily gets $a_n(u) \to 0$, which yields $I_n^{(1)} \to 0$ as $n \to +\infty$. At the same time,

$$b_n(u) = C_n^{-1} [\nu_u(LF) - \nu_{n,u}(LF)] + (C_n^{-1} - C_n^{-1}) \nu_u(LF).$$

(7.54)

Since $C_n \to C > 0$, to prove $b_n(u) \to 0$ as $n \to +\infty$ by (7.54) it is enough to show that $\nu_u(LF) - \nu_{n,u}(LF) \to 0$ for $F \in \mathcal{D}(L)$. To this end we recall that $G = G_{m-1}(F_m \circ \varpi_{s_m})$, see (3.17). Set

$$\tilde{\nu}_{n,u}(A) = \tilde{C}_{n}^{-1} P_{s,u}^{\alpha_n}(G_{m-1}(1_A \circ \varpi_{s}), \quad u \in [s_m, s_m],$$

$$\tilde{\nu}_u(A) = \tilde{C}_{n}^{-1} P_{s,u}^{\alpha_n}(G_{m-1}(1_A \circ \varpi_{s})).$$

As above, we have that $\tilde{\nu}_{n,u}, \tilde{\nu}_u \in \mathcal{P}_\text{exp}^0$ for all $n$ and $u$ as above. Clearly, we may assume that $\vartheta > \tilde{\vartheta}$, and hence their correlation functions, $\tilde{k}_u$ and $\tilde{k}_u^{\alpha_n}$, lie in the corresponding $K_{\vartheta}$. As in (6.7) we then can write

$$k_u - k_u^{\alpha_n} = Q_{\vartheta}(u - s_m)\tilde{k}_{s_m} - Q_{\vartheta}(u - s_m)\tilde{k}_{s_m}^{\alpha_n},$$

(7.55)

For $m = 1$, $\tilde{k}_u$ and $\tilde{k}_u^{\alpha_n}$ are the correlation functions of $\mu_u$ and $\mu_u^{\alpha_n}$, and hence one may apply Lemma 6.3, which yields

$$\mu_u(LF) - \mu_u^{\alpha_n}(LF) = \langle \tilde{k}_u - \tilde{k}_u^{\alpha_n}, \tilde{L}G \rangle = \langle \tilde{k}_u - \tilde{k}_u^{\alpha_n}, G \rangle \to 0, \quad n \to +\infty,$$

where $\tilde{G} \in \cap \mathcal{G}_{\vartheta}$ is such that $F = K\tilde{G}$, see (5.1) and (5.2). Therefore, we may inductively assume in (7.55) that

$$\langle \tilde{k}_{s_m} - \tilde{k}_{s_m}^{\alpha_n}, \tilde{L}G \rangle \to 0,$$

and obtain

$$\nu_u(LF) - \nu_{n,u}(LF) = \langle k_u - k_u^{\alpha_n}, \tilde{L}G \rangle \to 0,$$

by repeating the steps made in the proof of Lemma 6.3. This yields $b_n(u) \to 0$. As already mentioned above, both terms of $b_n(u)$ are bounded uniformly in $n$ and $u$, that yields in (7.53) $I_n^{(2)} \to 0$.

Let us now turn to $I_n^{(3)}$. As above, we have here

$$|c_n(u)| = \left| \langle \tilde{k}_u^{\alpha_n}, \tilde{L}_nG \rangle \right| \leq e^\vartheta |\tilde{L}_nG|_{\vartheta},$$

where $G \in \cap \mathcal{G}_{\vartheta}$ is such that $F = K\tilde{G}$, see (5.1) and (5.2), and $\tilde{L}_n$ is obtained by replacing $a(x-y)$ in (4.1) by

$$a_n(x,y) = a(x-y)(1 - \psi_{s_m})(x) = \alpha_n a(x-y)|x|^{d+1} \leq |x|^{d+1} = \alpha_n a(x-y)|x|^{d+1} =: \alpha_n a(x,y).$$

Proceeding as in obtaining (4.5) we then get, see (3.3),

$$|\tilde{L}_nG|_{\vartheta} \leq \frac{2\alpha_n m_{d+1}}{c(\vartheta - \vartheta')} \exp(e^\vartheta |\phi|) |G|_{\vartheta'}.$$

Here $\vartheta'$ can be an arbitrary number since $G \in \cap \mathcal{G}_{\vartheta'}$, see (5.2). This yields $I_n^{(3)} \to 0$ as $n \to +\infty$ (and hence $\alpha_n \to 0$), which by (7.53) implies (7.52). Therefore, the proof of Theorem 6.1 is completed.
7.4 Proof of Theorem 3.6

Claim (a) follows by Theorem 6.1 (existence) and Theorem 5.2 (uniqueness). The validity of (b) is then a standard fact, cf. [13, Theorem 5.1.2, claim (iv), page 80]. To prove (c), we proceed as follows. By construction, the law of \( X(t) \) is \( \mu_t \in \mathcal{P}_\text{exp} \); hence, \( X(t) \in \Gamma_* \) with probability one, see Lemma 2.10. Let \( \{D_k\}_{k \in \mathbb{N}} \) and \( \{\Gamma_{*,k}\}_{k \in \mathbb{N}} \) be the collections of balls and sets, respectively, used in the proof of Lemma 2.7, see (2.27). As we show there, each \( \Gamma_{*,k} \) is an open subset of \( \Gamma_* \), and \( \Gamma_* = \bigcap_k \Gamma_{*,k} \). For \( H \) as in (2.31) and \( k \in \mathbb{N} \), we set \( H_k(\gamma) = H(\gamma_k) = H(\gamma \cap D_k) \). Then \( H_k(\gamma) < \infty \) for \( \gamma \in \Gamma_{*,k} \). For \( N \in \mathbb{N} \) and \( s \geq 0 \), let us consider the following stopping time

\[
T^k_N = \inf\{ t \geq s : H_k(X(t)) > N \},
\]

cf. [15, page 180], and then set \( T^k_N \wedge t = \min\{T^k_N; t\} \) and \( Z(t) = \lim_{N \to +\infty} X(T^k_N \wedge t) \), which exists as \( T^k_N \leq T^k_{N+1} \). Let \( \Phi^m_{\tau} \in \mathcal{D}(L) \) be the same as in (5.33). Then

\[
\Phi^m_{\tau}(X(t)) - \int_s^t (L\Phi^m_{\tau})(X(u))du
\]
is a right-continuous martingale. Let \( \tilde{\mu}_t \) be the law of \( Z(t) \) and \( T^k = \lim_{N \to +\infty} T^k_N \). Similarly as in [15, page 180], for each \( t > s \) by the optional sampling theorem, we can write

\[
E[\Phi^m_{\tau}(X(T^k_N \wedge t))] = E[\Phi^m_{\tau}(X(s))] + E\left[ \int_s^{T^k_N \wedge t} (L\Phi^m_{\tau})(X(u))du \right],
\]

which after passing to the limit \( N \to +\infty \) yields

\[
\tilde{\mu}_t(\Phi^m_{\tau}) = \mu(\Phi^m_{\tau}) + E\left[ \int_s^{T^k_N \wedge t} (L\Phi^m_{\tau})(X(u))du \right]
\]

\[
\leq \mu(\Phi^m_{\tau}) + E\left[ \int_s^{T^k_N \wedge t} |(L\Phi^m_{\tau})(X(u))| \right]
\]

\[
\leq \mu(\Phi^m_{\tau}) + E\left[ \int_s^t |(L\Phi^m_{\tau})(X(u))| \right] \leq \mu(\Phi^m_{\tau}) + \int_s^t \mu_u(\Phi^m_{\tau,1})du,
\]

where \( \mu_u = P_{s,u} \circ \phi_u^{-1} \) is the law of \( X(u) \). Note that in the last line we used (5.24). By this estimate and (5.35) (with \( \theta_0 = \mu \) we then get the following

\[
\tilde{\mu}_t(\Phi^m_{\tau}) \leq \sum_{n=0}^{+\infty} \frac{(t-s)^n}{n!} \mu(\Phi^m_{\tau,n}).
\]

Now we proceed as in (5.38) and arrive at

\[
\lim_{\tau \to 0} \tilde{\mu}_t(\Phi^m_{\tau}) \leq (\kappa e^{2(t-s)})m ||\theta||^m_{L^1(X)},
\]

holding for all \( m \in \mathbb{N} \) and \( t-s < \rho_c \). Here \( \kappa \) is the type of \( \mu \). This yields that \( \tilde{\mu}_t \in \mathcal{P}_{\text{exp}} \) for such \( t \), and hence \( Z(t) \in \Gamma_* \) almost surely, implying \( T^k > t \). Now we fix \( v < s + \rho_c \), repeat this procedure with the martingale

\[
\Phi^m_{\tau}(X(t + v)) - \int_{s+v}^{t+v} (L\Phi^m_{\tau})(X(u))du
\]

and eventually conclude that \( T^k > t \) for all \( t \), and hence almost all sample paths of \( X \) remain in \( \mathcal{D}_{[s,\infty)}(\Gamma_{*,k}) \), holding for every \( k \). Since \( \Gamma_* = \bigcap_k \Gamma_{*,k} \), this yields that these paths remain in \( \mathcal{D}_{[s,\infty)}(\Gamma_*) \), cf. [15, Proof of Proposition 3.10, page 180, 181], which complete the proof.
Appendix. Supplementary Material

Here we prove (5.24), (5.25) and (5.26). For \( n = 2 \), (5.24) is just (5.20) with \( \Phi^m_{\tau,2} \) given by (5.25) with \( w_1(m, 2) = (m + 1)^2 - m^2 = 2m + 1 \), \( w_2(m, 2) = 1 \), see (5.28) and (5.27). Assume then that \( \Phi^m_{\tau,n} \) is as in (5.25). By (5.23), similarly as in (3.12), we get

\[
|LV_{\tau}(c; \gamma)| \leq \sum_{j=1}^{m} \hat{c}_m \theta^{\gamma_j}, \ldots, \theta^{\gamma_{j+1}} \theta^{\gamma_{j+1} + 1}, \ldots, \theta^{\gamma_m} \theta^{\gamma_m} + \tau c_a \hat{F}_{\tau}^{n+1}(\gamma).
\]

Here we have taken into account that \( \hat{c}_0 = 1 \), and also \( \theta^\gamma \leq c_a^\gamma \psi(x) \), see (3.6) and (3.7). In a similar way, by (3.13) we obtain

\[
|L \hat{F}_{\tau}^{m}(\gamma)| \leq mc_a \hat{F}_{\tau}^{m}(\gamma) + \tau c_a \hat{F}_{\tau}^{n+1}(\gamma).
\]

Now we use both this estimates in (5.25) and obtain

\[
|L \Phi^m_{\tau,n}(\gamma)| \leq \sum_{c \in C_{m,n}} C_{m,n}(c) \left( c_0 V_{\tau}(c_0 - 1, c_1 + 1, c_2, \ldots, c_k; \gamma) + c_1 V_{\tau}(c_0, c_1, c_2, \ldots, c_k; \gamma) + \ldots + c_n V_{\tau}(c_0, c_1, \ldots, c_n - 1, c_{n+1}; \gamma) \right) + \tau c_a \hat{F}_{\tau}^{n+1}(\gamma) + c_a \sum_{k=1}^{n} \tau^{k} (m + k) w_k(m, n) \hat{F}_{\tau}^{n+k}(\gamma) + \sum_{k=2}^{n+1} \tau^{k} (m + k) w_k(m, n) \hat{F}_{\tau}^{n+k}(\gamma).
\]

If one takes into account the recurrence formulas in (5.27), the latter two lines of the right-hand side of (7.56) convert into the second term of (5.25) written for \( \Phi^m_{\tau,n+1} \). Thus, it remains to prove that the first three lines of (7.56) yield the first term of (5.25) written for \( \Phi^m_{\tau,n+1} \). Note that therein the summands corresponding to \( c_j = 0 \) vanish automatically since we multiply them by zero in this case. Assuming that a given \( c_j \neq 0 \) we can write the corresponding summand in (7.56), denoted \( S_{n+1} \), as follows, see the second line in (5.25).

\[
S_{n+1} = \frac{m!n!(j + 1)!((c_{j+1} + 1) - (c_j - 1)/(c_{j+1} + 1)! \cdots \cdots (j!)^{c_j} - 1!/((j + 1)!))^{c_{j+1}+1} \cdots (c_{j+1} + 1, \ldots, c_j)}{c_0, c_1, \ldots, (c_j - 1), (c_{j+1} + 1) \cdots \cdots \cdots \cdots (j!)^{c_j} - 1!} W_{\tau}(c; \gamma), \quad c' \in C_{m,n+1},
\]

where \( c' = (c_0, \ldots, c_j - 1, c_{j+1} + 1, \ldots) \). To get convinced that \( c' \) is indeed in \( C_{m,n+1} \) one computes the corresponding sums, cf. (5.22), that yields \( c_0 + \cdots + c_j - 1 + c_{j+1} + 1 + \cdots = c_0 + \cdots + c_j + c_{j+1} + \cdots = m \), \( c_1 + \cdots + j(c_j - 1) + (j + 1)(c_{j+1} + 1) + \cdots = c_1 + \cdots + j\gamma_j + (j + 1)c_{j+1} + \cdots + (j + 1) = n + 1 \). Then we rewrite each summand in the first three lines of (7.56) as in (7.57) and observe that the corresponding \( c' \) runs over the whole \( C_{m,n+1} \) when \( c \) runs through \( C_{m,n} \). Then these three lines, denoted \( S_{n+1} \), take
A Markov process for a particle system

the following form

\[
S^{n+1} = \sum_{c' \in C_{m,n+1}} \left( \frac{1}{n+1} \sum_{j=1}^{n+1} j c'_j \right) C_{m,n+1}(c') V_{\tau}(c'; \gamma) \\
= \sum_{c' \in C_{m,n+1}} C_{m,n+1}(c') V_{\tau}(c'; \gamma),
\]

where we have taken into account that \( \sum_j j c_j = n + 1 \), see (5.22). This completes the proof of (5.24) and (5.25). It then remains to prove (5.26). For \( n = 1 \), \( C_{m,1} \) is a singleton consisting of \( c = (m-1, 1, 0, \ldots) \), which yields

\[
\sum_{c \in C_{m,1}} C_{m,1}(c) = \frac{m!}{(m-1)!1!} = m.
\]

Now we set in the second line of (7.58) \( V_{\tau}(c'; \gamma) = 1 \) and calculate \( S^{n+1} \) with this \( V_{\tau} \), which is equal to the first three lines of (7.56). That is,

\[
\sum_{c' \in C_{m,n+1}} C_{m,n+1}(c') = \sum_{c \in C_{m,n}} C_{m,n}(c) \left( c_0 + c_1 + \cdots + c_n \right) = m \sum_{c \in C_{m,n}} C_{m,n}(c),
\]

where we once again have used the first equality in (5.22). Now (5.26) is obtained from the latter by the induction in \( n \).

References


EJP 26 (2021), paper 72.
A Markov process for a particle system


Acknowledgments. Yuri Kozitsky thanks Lucian Beznea, Oleh Lopushansky and Yuri Tomilov for discussing some of its aspects. He thanks also Lucian Beznea and BIT DEFENDER for warm hospitality and financial support during his stay in Bucharest in April 2019, where a part of this work was done. Last but not least, the authors are cordially grateful to the referee for valuable and favorable suggestion that helped to improve the quality of the final version of this work.
Advantages of publishing in EJP-ECP

- Very high standards
- Free for authors, free for readers
- Quick publication (no backlog)
- Secure publication (LOCKSS\textsuperscript{1})
- Easy interface (EJMS\textsuperscript{2})

Economical model of EJP-ECP

- Non profit, sponsored by IMS\textsuperscript{3}, BS\textsuperscript{4}, ProjectEuclid\textsuperscript{5}
- Purely electronic

Help keep the journal free and vigorous

- Donate to the IMS open access fund\textsuperscript{6} (click here to donate!)
- Submit your best articles to EJP-ECP
- Choose EJP-ECP over for-profit journals

\textsuperscript{1}LOCKSS: Lots of Copies Keep Stuff Safe \url{http://www.lockss.org/}
\textsuperscript{2}EJMS: Electronic Journal Management System \url{http://www.vtex.lt/en/ejms.html}
\textsuperscript{3}IMS: Institute of Mathematical Statistics \url{http://www.imstat.org/}
\textsuperscript{4}BS: Bernoulli Society \url{http://www.bernoulli-society.org/}
\textsuperscript{5}Project Euclid: \url{https://projecteuclid.org/}
\textsuperscript{6}IMS Open Access Fund: \url{http://www.imstat.org/publications/open.htm}