

Bilinear Coagulation via Random Graphs

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Motivation: Interaction Clusters in Kac Dynamics

Starting Point: the Boltzmann-Grad Limit

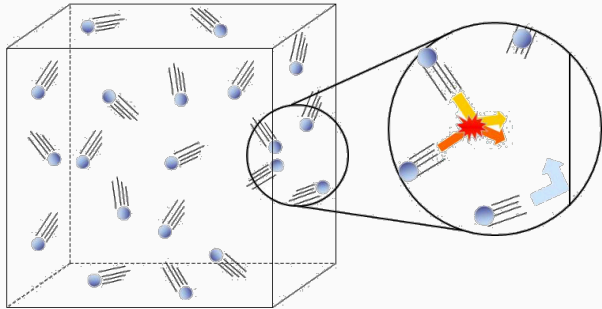
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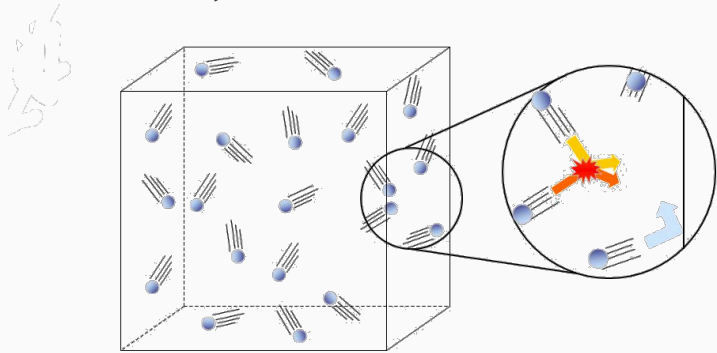
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- Derivation of the **spatially inhomogeneous Boltzmann equation** from molecular dynamics:



- Ballistic dynamics, hard core exclusion of radius $r_N \sim N^{-1/2}$

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 - Bogolyubov's *interaction clusters* partition N particles.
 - Related to short-time proof of the Boltzmann equation by Simonella, Gallagher, Bodineau: Two particles are in the same interaction cluster if the associated collision trees are joined by a recollision.

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 - When particles collide, energy and momentum are conserved.

From Interaction Clusters to Bilinearity and Random Graphs

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- If $\Psi(|v|) = a + b|v|^2$, then $K(x, y, S)$ computes

$$\begin{aligned} K(x, y, S) &= \sum_{v \in x, w \in y} (a + b|v - w|^2) = \sum_{v \in x, w \in y} (a + b|v|^2 - 2v \cdot w + |w|^2) \\ &= aN(x)N(y) + bE(x)N(y) - 2bP(x) \cdot P(y) + bN(x)E(y) \end{aligned} \tag{1}$$

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- π is **additive**: $\pi(z) = \pi(x) + \pi(y)$ for $K(x, y, \cdot)$ -a.e. z , $\pi(x') = \pi(x)$ for $J^N(x, \cdot)$ -a.e. x' .

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- Merger rates don't depend on scattering angles, so **projected** empirical measure $\pi_{\#}\mu_t^N = N^{-1} \sum \delta_{\pi(x_i^N(t))}$ has the same rates if the 'post-collisional' velocities are the same as the incoming!

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- "Any fact about μ_t^N which depends only on the conserved quantities is the same for the graph model".

Coagulation Equations: General Framework

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- Smolouchowski equation, weak form:

$$\frac{d}{dt} \langle f, \mu_t \rangle = \underbrace{\int_{S \times S \times S} (f(z) - f(x) - f(y)) K(x, y, dz) \mu_t(dx) \mu_t(dy)}_{=: \langle f, L(\mu_t) \rangle}$$

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for $f \in C_c(S)$.

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- Smolouchowski equation treats the gel as inert, but this is not the case for the microscopic dynamics.
- Flory equation, accounting for the gel $g_t = \langle\pi, \mu_0 - \mu_t\rangle \geq 0$:

$$\frac{d}{dt}\langle f, \mu_t\rangle = \langle f, L(\mu_t)\rangle - \int_{S \times S} f(x)\bar{K}(x, y)\mu_t(dx)(\mu_0 - \mu_t)(dy). \quad (2)$$

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- Gelation time

$$t_{\text{gel}} := \inf\{t : g_t \neq 0\}.$$

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4. g_t is differentiable except at t_{gel} , right-differentiable at the gelation time t_{gel} and $g'_{t_{\text{gel}}+} > 0$. Moreover, for some convex combination $\theta_i \geq 0, \sum_{i=1}^n \theta_i = 1$, it holds that

$$\frac{\sum_{i=1}^n \theta_i (g'_{t_{\text{gel}}+})_i}{(g'_{t_{\text{gel}}})_0} > \sum_{i=1}^n \frac{\theta_i \langle \pi_i, \mu_0 \rangle}{\langle \pi_0, \mu_0 \rangle}.$$

Main Result: Convergence of the Particle System

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- The empirical measure $\mu_t^N := N^{-1} \sum \delta_{x_i^N(t)} \rightarrow \mu_t$, where μ_t is the unique solution to the Flory equation starting at μ_0 , uniformly in $t \geq 0$, in probability and

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- In particular, t_{gel} is the first time a macroscopic cluster appears, and this cluster contains all of the conserved quantities escaping to infinity.

Key Elements in the Proof

Overview of the Proof

1. Existence and Uniqueness of Flory Equation via truncated equations on $S_\xi := \{x \in S : \varphi(x) \leq \xi\}$.

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4. Identification of t_{gel} with

$$t_{\text{expl}} := \inf\{t \geq 0 : \mathcal{E} \notin L^\infty([0, t])\}.$$

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3. Convergence of the gel $g_t^N \rightarrow g_t$, which implies $t_{\text{gel}} = t_{\text{crit}}$ and gives the characterisation of t_{gel} .
4. Identification of t_{gel} with

$$t_{\text{expl}} := \inf\{t \geq 0 : \mathcal{E} \notin L^\infty([0, t])\}.$$

5. Finiteness of \mathcal{E} on (t_{gel}, ∞) by duality argument.

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- Conclude with diagonal argument and tuning ξ_N .

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- Construct ν_0^ϵ by shifting ν_0 by $(\epsilon, \epsilon, \dots, \epsilon)$ and cutting out large values.

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- Characterisations of both blowup and gelation times are now continuous in the initial data, so we can take limits of
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- As a result, $\mu_t = \hat{\mu}_t$, which has finite second moments thanks to the behaviour **before** gelation!

Back to Interaction clusters

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- **So: verified, for toy model, that macroscopic interaction clusters occur before t_{mf} !**