Oscillatory kinetics in cluster-cluster aggregation

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Aggregation phenomena : motivation



- Many particles of one material dispersed in another.
- Transport is diffusive or advective.
- Particles stick together on contact.

Applications: surface physics, colloids, atmospheric science, earth sciences, polymers, cloud physics.

This talk:

Today we will focus on simple theoretical models of the statistical dynamics of such systems.

Simplest model of clustering: coalescing random walks



Cartoon of dynamics in 1-D with $k \to \infty$.

- Particles perform random walks on a lattice.
- Multiple occupancy of lattice sites is allowed.
- Particles on the same site merge with probability rate k: A + A → A.
- A source of particles may or may not be present.

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• No equilibrium - lack of detailed balance.

Mean field description

Equation for the average density, N(x, t), of particles:

$$\partial_t N = D \Delta N - k N^{(2)} + J$$

k - reaction rate, *D* - diffusion rate, *J* - injection rate, $N^{(2)}$ - density of same-site pairs.

Mean field assumption:

- No correlations between particles: $N^{(2)} \sim \frac{1}{2} N^2$:
- Spatially homogeneous case, N(x, t) = N(t).

Mean field rate equation:

$$\frac{dN}{dt} = -\frac{1}{2} k N^2 + J \qquad N(0) = N_0$$

$$J = 0: N(t) = \frac{2N_0}{2+kN_0 t} \sim \frac{1}{kt} \text{ as } t \to \infty$$

$$J \neq 0: N(t) \sim \sqrt{\frac{2J}{k}} \text{ as } t \to \infty$$

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A more sophisticated model of clustering: size-dependent coalescence

A better model would track the sizes distribution of the clusters:

$$A_{m_1}+A_{m_2}\to A_{m_1+m_2}.$$

- Probability rate of particles sticking should be a function, $K(m_1, m_2)$, of the particle sizes (bigger particles typically have a bigger collision cross-section).
- Micro-physics of different applications is encoded in K(m₁, m₂)
- Given the kernel, objective is to determine the cluster size distribution, N_m(t), which describes the average number of clusters of size m as a function of time.

Mean-field theory of irreversible coagulation

Assume the system is statistically homogeneous and well-mixed so that there are no spatial correlations. Particle size distribution, $N_m(t)$, satisfies the kinetic equation :

Smoluchowski equation :

$$\partial_t N_m(t) = \frac{1}{2} \int_0^m dm_1 K(m_1, m - m_1) N_{m_1}(t) N_{m - m_1}(t) \\ - N_m(t) \int_0^M dm_1 K(m, m_1) N_{m_1}(t) \\ + J \delta(m - m_0)$$

Microphysics is encoded in the coagulation kernel, $K(m_1, m_2)$.

- Source: particles of size *m*₀ are continuously added to the system at rate *J*.
- Sink: particles larger than cut-off, *M*, are removed from the the system.

Scale-invariant coagulation kernels

Notation: In many applications kernel is homogeneous:

$$K(am_1, am_2) = a^{\lambda} K(m_1, m_2)$$

 $K(m_1, m_2) \sim m_1^{\mu} m_2^{\nu} m_1 \ll m_2.$

Clearly $\lambda = \mu + \nu$. Examples:

Brownian coagulation of spherical droplets ($\nu = \frac{1}{3}, \mu = -\frac{1}{3}$):

$$K(m_1, m_2) = \left(\frac{m_1}{m_2}\right)^{\frac{1}{3}} + \left(\frac{m_2}{m_1}\right)^{\frac{1}{3}} + 2$$

Gravitational settling of spherical droplets in laminar flow ($\nu=\frac{4}{3},\,\mu=$ 0) :

$$K(m_1, m_2) = \left(m_1^{\frac{1}{3}} + m_2^{\frac{1}{3}}\right)^2 \left|m_1^{\frac{2}{3}} - m_2^{\frac{2}{3}}\right|$$

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Self-similar solutions of Smoluchowski equation



For homogeneous kernels, cluster size distribution often selfsimilar. Without source:

$$N_m(t) \sim s(t)^{-2} F(z) \quad z = rac{m}{s(t)}$$

s(t) is the typical cluster size. The scaling function, F(z), determining the shape of the cluster size distribution, satisfies:

$$-2F(z) + z \frac{dF(z)}{dz} = \frac{1}{2} \int_0^z dz_1 K(z_1, z - z_1) F(z_1) F(z - z_1)$$

- $F(z) \int_0^\infty dz_1 K(z, z_1) F(z_1).$

Stationary solutions of the Smoluchowski equation with a source and sink



- Add monomers at rate, *J*. Remove those with *m* > *M*.
- Stationary state is obtained for large *t* which balances injection and removal.
- Constant mass flux in range [*m*₀, *M*]
- Model kernel:

$$K(m_1, m_2) = \frac{1}{2}(m_1^{\mu}m_2^{\nu} + m_1^{\nu}m_2^{\mu})$$

Stationary state for $t \to \infty$, $m_0 \ll m \ll M$ (Hayakawa 1987):

$$N_m = \sqrt{\frac{J(1 - (\nu - \mu)^2)\cos((\nu - \mu)\pi/2)}{2\pi}} m^{-\frac{\nu + \mu + 3}{2}}.$$
Require mass flux to be *local*: $|\mu - \nu| < 1$. But what if it is 12^{12} .

Asymptotic solution of the nonlocal case

Nonlocal stationary state is not like the Hayakawa solution:



Stationary state (theory vs numerics).

 Stationary state has the asymptotic form for *M* ≫ 1:

$$N_m = \frac{\sqrt{2J\gamma \log M}}{M} M^{m^{-\gamma}} m^{-\nu}.$$

$$\gamma = \nu - \mu - \mathbf{1}.$$

- Stretched exponential for small *m*, power law for large *m*.
- Agrees well with numerics without any adjustable parameters.

Note: the stationary state **vanishes** as $M \to \infty$. What happens version of the mass flux?

Hopf bifurcation of stationary state as M increased



Linear stability analysis

- We did a (semi-analytic) linear stability analysis of the exact stationary state.
- Concluded that the nonlocal stationary state is linearly unstable for large enough *M*.
- Constant mass flux is replaced by time-periodic pulses.
- Oscillatory behaviour due to an attracting limit cycle embedded in this very high-dimensional dynamical system.

Scaling of period and amplitude of oscillation with M



 Oscillations are a sequence of "resets" of self-similar pulses:

 $N_m(t) = s(t)^a F(\xi)$ with $\xi = \frac{m}{s(t)}$,

where

$$a = -rac{\nu + \mu + 3}{2}$$
 $s(t) \sim t^{rac{2}{1 - \nu - \mu}}.$

Period estimated as the time, *τ_M*, required for *s*(*τ_M*) ≈ *M*. Amplitude, *A_M*, estimated as the mass supplied in time *τ_M*:

$$au_M \sim M^{rac{1-
u-\mu}{2}}$$

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Other examples: collisional evaporation / fragmentation models

- With probability $\lambda \ll 1$, collisions result in:
 - evaporation (both particles removed) with *J* fixed.
 - complete fragmentation (both particles converted to monomers) with J = 0
- Rate equations are almost the same (except for equation for monomer density in fragmentation case):

$$\frac{\partial N_m(t)}{\partial t} = \frac{1}{2} \sum_{m_1=1}^m K_{m-m_1,m_1} N_{m-m_1} N_{m_1} - (1+\lambda) N_m \sum_{m_1=1}^\infty K_{m,m_1} N_{m_1} + J \,\delta_{m,1}$$

Keep the model kernel $K(m_1, m_2) = \frac{1}{2}(m_1^{\mu}m_2^{\nu} + m_1^{\nu}m_2^{\mu}).$

C. Connaughton, A. Dutta, R. Rajesh, and O. Zaboronski. Universality properties of steady driven coagulation with WICK collisional evaporation. EPL Europhys. Lett., 117(1):10002, 2017

Oscillatory kinetics for collisional evaporation / fragmentation model

Oscillatory regime observed for both models when $\lambda \ll 1$ (for certain kernels).



Oscillations in collisional evaporation / fragmentation models

Oscillations are not a result of "hard" cut-off.



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Summary:

- Stationary solution of Smoluchowski equation with source investigated in regime |ν – μ| > 1.
- Size distribution can be calculated asymptotically and has a novel functional form.
- Amplitude of state vanishes as the dissipation scale grows.
- Stationary state can become unstable so the long-time behaviour of the cascade dynamics is oscillatory.

Questions:

- Do any physical systems really behave like this?
- What happens in spatially extended systems?
- Other examples of oscillatory kinetics? Eg in wave turbulence?

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Violation of mass conservation: the gelation transition

Microscopic dynamics conserve mass: $A_{m_1} + A_{m_2} \rightarrow A_{m_1+m_2}$.



 Smoluchowski equation formally conserves the total mass,

$$M_1(t) = \int_0^\infty m N(m, t) \, dm.$$

• However for $\lambda > 1$:

$$M_1(t) < \int_0^\infty m N(m,0) \, dm \, t > t^*.$$

(Lushnikov [1977], Ziff [1980])

 Mean field theory violates mass conservation!!!

Best studied by introducing cut-off, *M*, and studying limit $M \rightarrow \infty$. (Laurencot [2004]) Physical interpretation? Intermediate asymptotics... Asymptotic behaviour of the kernel controls the aggregation of small clusters and large:

$$K(m_1,m_2) \sim m_1^{\mu}m_2^{\nu} m_1 \ll m_2.$$

 $\mu + \nu = \lambda$ so that gelation always occurs if ν is big enough.

Instantaneous Gelation

- If $\nu > 1$ then $t^* = 0$. (Van Dongen & Ernst [1987])
- Worse: gelation is *complete*: $M_1(t) = 0$ for t > 0.

Instantaneously gelling kernels cannot describe even the intermediate asymptotics of any physical problem. Mathematically pathological?

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Droplet coagulation by gravitational settling: a puzzle



- The process of gravitational settling is important in the evolution of the droplet size distribution in clouds and the onset of precipitation.
- Droplets are in the Stokes regime → larger droplets fall faster merging with slower droplets below them.

Some elementary calculations give the collision kernel

$$K(m_1, m_2) \propto (m_1^{\frac{1}{3}} + m_2^{\frac{1}{3}})^2 \left| m_1^{\frac{2}{3}} - m_2^{\frac{2}{3}} \right|$$

 $\nu = 4/3$ suggesting instantaneous gelation but model seems reasonable in practice. How is this possible?

Instantaneous gelation in the presence of a cut-off



- With cut-off, M, regularized gelation time, t_M^* , is clearly identifiable.
- t_M^* decreases as M increases.
- Van Dongen & Ernst recovered in limit *M* → ∞.
- Decrease of t_M^* as *M* is very slow. Numerics and heuristics suggest:

$$t^*_M \sim rac{1}{\sqrt{\log M}}.$$

This suggests such models are physically reasonable.

 Consistent with related results of Krapivsky and Ben-Naim and Krapivsky [2003] on exchange-driven growth.