

Asymptotic preserving schemes for nonlinear kinetic equations leading to volume-exclusion chemotaxis in the diffusive limit

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Outline

- 1 Introduction
- 2 Macro and kinetic models
- 3 Numerical methods for the kinetic model
- 4 Numerical experiments in 1D and 2D

Motivation

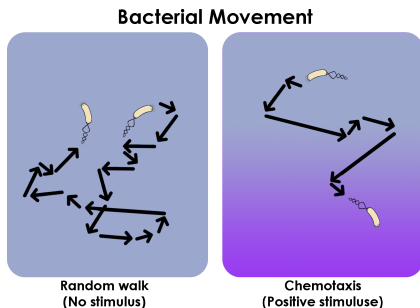


Figure: Cell movement without/with chemotaxis ¹

- **run and tumble**: cells move along a straight line in the running phase and reorient in the tumbling phase
- **chemotaxis**: cells adapt movement in response to a chemical stimulus

¹http://2016.igem.org/Team:Technion_Israel/Chemotaxis

Patlak-Keller-Segel (PKS) model

A popular macroscopic model:

$$\begin{aligned}\partial_t \rho - D_\rho \Delta \rho + \chi_0 \nabla \cdot \{\rho \nabla c\} &= h(\rho, c), \\ D_c \Delta c + g(\rho, c) &= 0.\end{aligned}$$

- ρ – cell density, c – chemical concentration
- D_ρ, D_c – diffusion constant
- $h(\rho, c) = r_0 \rho (1 - \rho / \rho_{\max})$ – proliferation or death of cells
- $g(\rho, c) = \rho - c$ – production or consumption of chemoattractant
- Simple yet able to demonstrate pattern forming for suitable kinetics
- Solutions have a tendency to exhibit **finite-time blow-up**
 - **Density-dependent** chemotactic sensitivity and motility

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1D discrete random walk ²

Notations:

- $u_i(t)$ – prob. at $i \in \mathbb{Z}$ at time t
- $c_i(t)$ – chemical concentration
- T_i^+ – prob. per unit time of one-step jump to $i + 1$
- T_i^- – prob. per unit time of one-step jump to $i - 1$

$$\frac{\partial u_i}{\partial t} = T_{i-1}^+ u_{i-1} + T_{i+1}^- u_{i+1} - (T_i^+ + T_i^-) u_i$$

²Kevin Painter and Thomas Hillen (2002)

Relation with classical PKS

Assume

$$T_i^{\pm} = \alpha + \frac{\beta}{2}(c_{i\pm 1} - c_i)$$

then

$$\frac{\partial u_i}{\partial t} = \alpha(u_{i+1} - 2u_i + u_{i-1}) - \frac{\beta}{2}((u_{i+1} + u_i)(c_{i+1} - c_i) - (u_i + u_{i-1})(c_i - c_{i-1}))$$

Relation with classical PKS

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With some scaling and taking the limit

$$\frac{\partial u}{\partial t} = D_u \frac{\partial^2 u}{\partial x^2} - \frac{\partial}{\partial x}(\beta u \frac{\partial c}{\partial x})$$

volume-filling model

Taking into account the volume limitations

$$T_i^\pm = q(u_{i\pm 1})(\alpha + \frac{\beta}{2}(c_{i\pm 1} - c_i)),$$

where

- $q(u)$ – prob. finding space at its neighbouring location
- $q(U_{\max})$ – finite capacity
- $q(u) \geq 0$ for $0 \leq u \leq U_{\max}$

volume-filling model

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- $q(U_{\max})$ – finite capacity
- $q(u) \geq 0$ for $0 \leq u \leq U_{\max}$

An example:

$$q(u) = 1 - \frac{u}{U_{\max}}$$

Volume-exclusion PKS

$$\begin{aligned} \frac{\partial u_i}{\partial t} = & \alpha(q_i u_{i+1} - (q_{i+1} + q_{i-1})u_i + q_i u_{i-1}) \\ & - \frac{\beta}{2}((q_i u_{i+1} + q_{i+1} u_i)(c_{i+1} - c_i) - (q_{i-1} u_i + q_i u_{i-1})(c_i - c_{i-1})) \end{aligned}$$

In the limit, we arrive at

$$\frac{\partial u}{\partial t} = D_u \left(q(u) \frac{\partial^2 u}{\partial x^2} - u \frac{\partial^2 q(u)}{\partial x^2} \right) - \frac{\partial}{\partial x} (\beta q(u) u \frac{\partial c}{\partial x})$$

or equivalent divergence form

$$\frac{\partial u}{\partial t} = \frac{\partial}{\partial x} \left(D_u(q(u) - q'(u)u) \frac{\partial u}{\partial x} - \beta q(u) u \frac{\partial c}{\partial x} \right)$$

Volume-exclusion PKS

$$\begin{aligned} \frac{\partial u_i}{\partial t} = & \alpha(q_i u_{i+1} - (q_{i+1} + q_{i-1})u_i + q_i u_{i-1}) \\ & - \frac{\beta}{2}((q_i u_{i+1} + q_{i+1} u_i)(c_{i+1} - c_i) - (q_{i-1} u_i + q_i u_{i-1})(c_i - c_{i-1})) \end{aligned}$$

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or equivalent divergence form

$$\frac{\partial u}{\partial t} = \frac{\partial}{\partial x} \left(D_u(q(u) - q'(u)u) \frac{\partial u}{\partial x} - \beta q(u) u \frac{\partial c}{\partial x} \right)$$

Well defined: $q(u) - q'(u)u > 0$ if $q'(u) \leq 0$

Framework of kinetic model

General form of the kinetic model:

$$\partial_t f + \mathbf{v} \cdot \nabla (F[\rho](t, \mathbf{x}, \mathbf{v}) f) = \lambda q(\rho) (-f + \rho T(\mathbf{v}, \rho, \nabla c)) .$$

- $f(t, \mathbf{x}, \mathbf{v})$ – density, $\rho(t, \mathbf{x}) = \int_V f(t, \mathbf{x}, \mathbf{v}) d\mathbf{v}$ with V unit sphere
- $c(t, \mathbf{x})$ – chemical concentration
- $q(\rho)$ – probability to find space at its neighbouring locations
- λ – constant turning rate
- $F[\rho](t, \mathbf{x}, \mathbf{v})$ – anisotropic transport due to density limitation
- $T(\mathbf{v}, \rho, \nabla c)$ – prob. of a velocity jump to velocity \mathbf{v} satisfying

$$\int_V T(\mathbf{v}, \rho, \nabla c) d\mathbf{v} = 1$$

Choice of $F[\rho]$ and T

- Assume cells only move to non-overcrowded regions

$$F[\rho](t, \mathbf{x}, \mathbf{v}) = q(\rho(t, \mathbf{x} + \mathbf{v})) .$$

- Assume cells only make a turn in directions where space is available

$$T(\mathbf{v}, \rho, \nabla c) = \tilde{c}(t, \mathbf{x}) \psi(\mathbf{v}, \nabla c) q(\rho(t, \mathbf{x} + \mathbf{v})) ,$$

where $\tilde{c}(t, \mathbf{x})$ is a normalisation factor.

Remark

The assumptions imply cells will only make a turn

- if they are not already trapped in a high density region*
- only in directions where the density is not too large*

Diffusion scaling

- Introduce characteristic speed s and characteristic length scale L
- Assuming the space scaling such that

$$\tau_{run} \ll \tau_{drift} \ll \tau_{diff}$$

where

- $\tau_{run} = \frac{1}{\lambda}$
- $\tau_{drift} = \frac{L}{s}$
- $\tau_{diff} \approx \frac{L^2 \lambda}{s^2}$ since the diffusion coefficient $D \approx \mathcal{O}(\frac{s^2}{\lambda})$
- Introduce $\varepsilon \ll 1$ such that

$$\tau_{run} = \mathcal{O}(1), \quad \tau_{drift} = \mathcal{O}\left(\frac{1}{\varepsilon}\right), \quad \tau_{diff} = \mathcal{O}\left(\frac{1}{\varepsilon^2}\right)$$

Scaled kinetic model

With the diffusion scaling, we get the dimensionless equation

$$\varepsilon^2 \partial_t f^\varepsilon + \varepsilon \mathbf{v} \cdot \nabla (F_\varepsilon[\rho](t, \mathbf{x}, \mathbf{v}) f^\varepsilon) = q(\rho)(-f^\varepsilon + \rho T_\varepsilon(\mathbf{v}, \rho, \nabla c))$$

where

$$F_\varepsilon[\rho](t, \mathbf{x}, \mathbf{v}) = q(\rho(t, \mathbf{x} + \varepsilon \mathbf{v}))$$

$$T_\varepsilon(\mathbf{v}, \rho, \nabla c) = \frac{\psi_\varepsilon(\mathbf{v}, \nabla c) q(\rho(t, \mathbf{x} + \varepsilon \mathbf{v}))}{\int_V \psi_\varepsilon(\mathbf{v}, \nabla c) q(\rho(t, \mathbf{x} + \varepsilon \mathbf{v})) d\mathbf{v}}$$

$$\psi_\varepsilon(\mathbf{v}, \nabla c) = \psi_0(\mathbf{v}) + \varepsilon \psi_1(\mathbf{v}, \nabla c)$$

To recover the volume-exclusion PKS, we make the assumptions:

- $\int_V \psi_0(\mathbf{v}) d\mathbf{v} = 1$, $\int_V \psi_1(\mathbf{v}, \nabla c) d\mathbf{v} = 0$
- $\int_V \mathbf{v} \psi_0(\mathbf{v}) d\mathbf{v} = 0$, $\psi_1(\mathbf{v}, \nabla c) = \phi(\mathbf{v}) \cdot \nabla c$

Macroscopic limit

Theorem (formal)

The limit $\varepsilon \rightarrow 0$ of f^ε is $f^0 = \rho(t, \mathbf{x})\psi_0(\mathbf{v})$, where ρ solves

$$\partial_t \rho - \nabla \cdot (D_0(q(\rho) - \rho q'(\rho)) \nabla \rho - \beta \rho q(\rho) \nabla c) = 0 ,$$

with the diffusion coefficient D_0 and the chemotactic sensitivity parameter β given by

$$D_0 = \langle (\mathbf{v} \otimes \mathbf{v}) \psi_0(\mathbf{v}) \rangle \quad \text{and} \quad \beta = \langle \mathbf{v} \otimes \phi(\mathbf{v}) \rangle .$$

where $\langle f(\mathbf{v}) \rangle = \int_V f(\mathbf{v}) \, d\mathbf{v}$.

General idea of proof: Consider a regular expansion of f^ε

$$f^\varepsilon(t, \mathbf{x}, \mathbf{v}) = f^0(t, \mathbf{x}, \mathbf{v}) + \varepsilon f^1(t, \mathbf{x}, \mathbf{v}) + \mathcal{O}(\varepsilon^2) .$$

Do expansions to the kinetic model and compare terms with the same order in ε .

Asymptotic expansion

- Expansion of $F_\varepsilon[\rho] = q(\rho(t, \mathbf{x} + \varepsilon \mathbf{v}))$:

$$F_\varepsilon[\rho] = q(\rho) + \varepsilon q'(\rho) \mathbf{v} \cdot \nabla \rho + \mathcal{O}(\varepsilon^2)$$

- Expansion of $T_\varepsilon = \tilde{c}_\varepsilon(t, \mathbf{x}) \psi(\mathbf{v}, \nabla c) q(\rho(t, \mathbf{x} + \varepsilon \mathbf{v}))$:

$$T_\varepsilon(\mathbf{v}, \rho, \nabla c) = \psi_0(\mathbf{v}) + \varepsilon \left(\psi_1(\mathbf{v}, \nabla c) + \frac{q'(\rho)}{q(\rho)} (\mathbf{v} \cdot \nabla \rho) \psi_0(\mathbf{v}) \right) + \mathcal{O}(\varepsilon^2)$$

Identifying equations in powers of ε , we obtain

$$\varepsilon^0 : f^0(t, \mathbf{x}, \mathbf{v}) = \rho(t, \mathbf{x}) \psi_0(\mathbf{v})$$

$$\varepsilon^1 : \mathbf{v} \cdot \nabla (q(\rho) f^0) = q(\rho) \left(-f^1 + \rho \left(\psi_1(\mathbf{v}, \nabla c) + \frac{q'(\rho)}{q(\rho)} (\mathbf{v} \cdot \nabla \rho) \psi_0(\mathbf{v}) \right) \right)$$

$$\varepsilon^2 : \partial_t f^0 + \mathbf{v} \cdot \nabla \left(q(\rho) f^1 + q'(\rho) (\mathbf{v} \cdot \nabla \rho) f^0 \right) = \mathcal{O}(1)$$

Formal proof of the limit $\varepsilon \rightarrow 0$

Replacing f^0 by $\rho(t, \mathbf{x})\psi_0(\mathbf{v})$, integrating ε^2 -equation over $\mathbf{v} \in V$

$$\partial_t f^0 + \mathbf{v} \cdot \nabla \left(q(\rho) f^1 + q'(\rho) (\mathbf{v} \cdot \nabla \rho) f^0 \right) = \mathcal{O}(1)$$

and noticing the right hand terms integrate to zero, we get

$$\partial_t \rho + \nabla \cdot (q(\rho) \langle \mathbf{v} f^1 \rangle + D_0 \rho q'(\rho) \nabla \rho) = 0, \quad (1)$$

where $D_0 = \langle (\mathbf{v} \otimes \mathbf{v}) \psi_0(\mathbf{v}) \rangle$.

Replacing f^0 by $\rho(t, \mathbf{x})\psi_0(\mathbf{v})$, multiplying ε -equation

$$\mathbf{v} \cdot \nabla (q(\rho) f^0) = q(\rho) \left(-f^1 + \rho \left(\psi_1(\mathbf{v}, \nabla c) + \frac{q'(\rho)}{q(\rho)} (\mathbf{v} \cdot \nabla \rho) \psi_0(\mathbf{v}) \right) \right)$$

by \mathbf{v} and integrating over $\mathbf{v} \in V$, we get

$$q(\rho) \langle \mathbf{v} f^1 \rangle = -\nabla \cdot (D_0 \rho q(\rho)) + \rho q(\rho) \langle \mathbf{v} \psi_1 \rangle + D_0 \rho q'(\rho) \nabla \rho \quad (2)$$

Combining (1) and (2), we recovered the volume-exclusion PKS model.

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Approximate kinetic model

As we are interested in the limit $\varepsilon \rightarrow 0$, we truncate F_ε and T_ε and solve

$$\varepsilon^2 \partial_t f + \varepsilon \mathbf{v} \cdot \nabla (\tilde{F}_\varepsilon(\rho) f) = q(\rho) \left(-f + \rho \tilde{T}_\varepsilon(\mathbf{v}, \rho, \nabla c) \right) + \varepsilon^2 r_0 f (1 - \rho/\rho_{\max})_+ ,$$

where

$$\tilde{F}_\varepsilon = q(\rho(t, \mathbf{x})) + \varepsilon q'(\rho(t, \mathbf{x})) \mathbf{v} \cdot \nabla \rho(t, \mathbf{x}) ,$$

$$\tilde{T}_\varepsilon = \psi_0(\mathbf{v}) + \varepsilon \left(\psi_1(\mathbf{v}, \nabla c) + \frac{q'(\rho)}{q(\rho)} (\mathbf{v} \cdot \nabla \rho) \psi_0(\mathbf{v}) \right) .$$

Remark

The approximate kinetic model converges to the same macroscopic limit (with proliferation term) as $\varepsilon \rightarrow 0$.

Micro-macro decomposition ³

To design an asymptotic preserving scheme, we decompose

$$f(t, \mathbf{x}, \mathbf{v}) = \rho(t, \mathbf{x})\psi_0(\mathbf{v}) + \varepsilon g(t, \mathbf{x}, \mathbf{v}),$$

where g is the perturbation satisfying $\langle g \rangle = 0$.

Equation for ρ

Integrating the approximate kinetic model

$$\partial_t f + \frac{1}{\varepsilon} \mathbf{v} \cdot \nabla (\tilde{F}_\varepsilon(\rho) f) = \frac{1}{\varepsilon^2} q(\rho) \left(-f + \rho \tilde{T}_\varepsilon(\mathbf{v}, \rho, \nabla c) \right) + r_0 f (1 - \rho/\rho_{\max})_+$$

over \mathbf{v} , we have

$$\partial_t \rho + \langle \mathbf{v} \cdot \nabla (q(\rho)g) \rangle + \nabla \cdot (q'(\rho)\rho D_0 \nabla \rho) = r_0 \rho (1 - \rho/\rho_{\max})_+.$$

³M. Lemou and L. Mieussens (2008)

Equation for g

To get equation for g , we introduce the projection

$$\Pi f(t, \mathbf{v}, \mathbf{x}) = \langle f(t, \mathbf{v}, \mathbf{x}) \rangle \psi_0(\mathbf{v})$$

Equation for g

Taking $I - \Pi$ into the kinetic equation, we get

$$\begin{aligned} & \partial_t g + \frac{1}{\varepsilon} (I - \Pi) \left[\mathbf{v} \cdot \nabla (q(\rho)g) + \nabla \cdot (\rho q'(\rho)(\mathbf{v} \otimes \mathbf{v}) \nabla \rho \psi_0) \right] \\ &= \frac{1}{\varepsilon^2} \left[-q(\rho)(\mathbf{v} \cdot \nabla \rho) \psi_0 + \rho q(\rho) \psi_1 - q(\rho)g \right] + r_0 g (1 - \rho/\rho_{\max})_+ . \end{aligned}$$

Micro-macro formulation

Micro-macro formulation

$$\begin{cases} \partial_t \rho + \langle \mathbf{v} \cdot \nabla (q(\rho)g) \rangle + D_0 \nabla \cdot (\rho q'(\rho) \nabla \rho) = r_0 \rho (1 - \rho/\rho_{\max})_+ , \\ \partial_t g + \frac{1}{\varepsilon} (I - \Pi) K_\varepsilon = \frac{1}{\varepsilon^2} S_\varepsilon + r_0 g (1 - \rho/\rho_{\max})_+ , \\ \Delta c + \rho - c = 0 , \end{cases}$$

where $D_0 = \langle (\mathbf{v} \otimes \mathbf{v}) \psi_0 \rangle$, and

$$\begin{aligned} K_\varepsilon &= \mathbf{v} \cdot \nabla (q(\rho)g) + \nabla \cdot (\rho \psi_0 (\mathbf{v} \otimes \mathbf{v}) \nabla q(\rho)) , \\ S_\varepsilon &= -q(\rho) (\mathbf{v} \cdot \nabla \rho) \psi_0 + \rho q(\rho) \psi_1 - q(\rho)g . \end{aligned}$$

Remark

In the limit $\varepsilon \rightarrow 0$, we can recover the volume-exclusion PKS model by forcing $S_\varepsilon = 0$.

1D finite difference scheme for ρ

Equation for ρ

With standard finite difference notations

$$\delta_t^+ \rho_j^n + \langle v_k \delta_x (q(\bar{\rho}_*^n) g_{*,k}^{n+1}) \rangle_j + D_h \delta_x (\bar{\rho}_*^n q'(\bar{\rho}_*^n) \delta_x \rho_*^{n+1})_j = r_0 \rho_j^n \left(1 - \frac{\rho_j^n}{\rho_{\max}} \right) +$$

where $\langle \eta_{j,k}^n \rangle_h := \Delta v \sum_k \eta_{j,k}^n$, $D_h := \langle v_k^2 \psi_0(v_k) \rangle_h$ and














$$\delta_x (q(\bar{\rho}_*^n) g_{*,k}^n)_j = \frac{q(\bar{\rho}_{j+\frac{1}{2}}^n) g_{j+\frac{1}{2},k}^n - q(\bar{\rho}_{j-\frac{1}{2}}^n) g_{j-\frac{1}{2},k}^n}{\Delta x},$$

Upwind discretization for chemotaxis

- Cells aggregate to form patterns in regions where chemoattractant is highly concentrated
- To avoid instability issues and inspired by ⁴, we discretize $q(\rho)\rho\psi_1(\mathbf{v}, \nabla c)$ in an **upwind** manner as follows

$$\Phi_{j+\frac{1}{2}}^{n_1, n} = \begin{cases} \rho_j^{n_1} q(\rho_{j+1}^n) , & \text{if } \delta_x c_{j+\frac{1}{2}}^n \geq 0 , \\ \rho_{j+1}^{n_1} q(\rho_j^n) , & \text{if } \delta_x c_{j+\frac{1}{2}}^n < 0 , \end{cases}$$

where $n_1 = n$ or $n + 1$.

⁴L. Almeida, F. Bubba, B. Perthame, and C. Pouchol. (2018)             

1D finite difference scheme for g

Equation for g

$$\delta_t^+ g_{j+\frac{1}{2},k}^n + \frac{1}{\varepsilon} (I - \Pi_h) K_{j+\frac{1}{2},k}^n = \frac{1}{\varepsilon^2} S_{j+\frac{1}{2},k}^{n,n+1} + r_0 g_{j+\frac{1}{2},k}^n \left(1 - \frac{\rho_{j+\frac{1}{2}}^n}{\rho_{\max}} \right)_+$$

where $\Pi_h \eta_{j,k}^n = \langle \eta_{j,k}^n \rangle_h \psi_0(v_k)$ for some general η and

$$K_{j+\frac{1}{2},k}^n = v_k^+ \delta_x (q(\bar{\rho}_*^n) g_{*,k}^n)_j - v_k^- \delta_x (q(\bar{\rho}_*^n) g_{*,k}^n)_{j+1} \\ + v_k^2 \psi_0(v_k) \delta_x (\bar{\rho}_*^n q'(\bar{\rho}_*^n) \delta_x \rho_*^n)_{j+\frac{1}{2}}$$

$$S_{j+\frac{1}{2},k}^{n,n+1} = -v_k \psi_0(v_k) q(\bar{\rho}_{j+\frac{1}{2}}^n) \delta_x \rho_{j+\frac{1}{2}}^{n+1} + \psi_1(v_k, \delta_x c_{j+\frac{1}{2}}^n) \Phi_{j+\frac{1}{2}}^{n+1,n} \\ - q(\bar{\rho}_{j+\frac{1}{2}}^n) g_{j+\frac{1}{2},k}^{n+1}$$

Decoupling of ρ and g

To decouple ρ and g , we define

$$\frac{\tilde{g}_{j+\frac{1}{2},k}^{n+1} - g_{j+\frac{1}{2},k}^n}{\Delta t} + \dots = \frac{1}{\varepsilon^2} \tilde{S}_{j+\frac{1}{2},k}^{n,n+1} + \dots,$$

where

$$\tilde{S}_{j+\frac{1}{2},k}^{n,n+1} = -v_k \psi_0(v_k) q(\bar{\rho}_{j+\frac{1}{2}}^n) \delta_x \rho_{j+\frac{1}{2}}^n + \psi_1(v_k, \delta_x c_{j+\frac{1}{2}}^n) \Phi_{j+\frac{1}{2}}^{n,n} - q(\bar{\rho}_{j+\frac{1}{2}}^n) \tilde{g}_{j+\frac{1}{2},k}^{n+1}.$$

Remark

$\tilde{g}_{j+\frac{1}{2},k}^{n+1}$ can be computed *explicitly*

Decoupling of ρ and g

To decouple ρ and g , we define

$$\frac{\tilde{g}_{j+\frac{1}{2},k}^{n+1} - g_{j+\frac{1}{2},k}^n}{\Delta t} + \dots = \frac{1}{\varepsilon^2} \tilde{S}_{j+\frac{1}{2},k}^{n,n+1} + \dots,$$

where

$$\tilde{S}_{j+\frac{1}{2},k}^{n,n+1} = -v_k \psi_0(v_k) q(\bar{\rho}_{j+\frac{1}{2}}^n) \delta_x \rho_{j+\frac{1}{2}}^n + \psi_1(v_k, \delta_x c_{j+\frac{1}{2}}^n) \Phi_{j+\frac{1}{2}}^{n,n} - q(\bar{\rho}_{j+\frac{1}{2}}^n) \tilde{g}_{j+\frac{1}{2},k}^{n+1}.$$

Remark

$\tilde{g}_{j+\frac{1}{2},k}^{n+1}$ can be computed *explicitly*

The equation for ρ can be reformulated as

$$\frac{\rho_j^{n+1}}{\Delta t} - \delta_x (a_*^n \delta_x \rho_*^{n+1})_j + \delta_x (b_*^n \Phi_*^{n+1,n})_j + D_h \delta_x (\rho_*^n q'(\rho_*^n) \delta_x \rho_*^{n+1})_j = r_j^n,$$

where $a_{j+\frac{1}{2}}^n$, $b_{j+\frac{1}{2}}^n$ and r_j^n can be *explicitly* computed using $\tilde{g}_{j+\frac{1}{2},k}^{n+1}$

Asymptotic preserving property

Checking the order of ε in the equation for g , we should have $S_{j+\frac{1}{2},k}^{n,n+1} = 0$ as $\varepsilon \rightarrow 0$, namely

$$q(\bar{\rho}_{j+\frac{1}{2}}^n)g_{j+\frac{1}{2},k}^{n+1} = -v_k\psi_0(v_k)q(\bar{\rho}_{j+\frac{1}{2}}^n)\delta_x\rho_{j+\frac{1}{2}}^{n+1} + \psi_1(v_k, \delta_x c_{j+\frac{1}{2}}^n)\Phi_{j+\frac{1}{2}}^{n+1,n}.$$

A direct computation shows that

$$\langle v_k \delta_x (q(\bar{\rho}_*^n)g_{*,k}^{n+1}) \rangle_j = \delta_x (-D_h q(\bar{\rho}_*^n) \delta_x \rho_*^{n+1} + \langle v_k \psi_1(v_k, \delta_x c_*^n) \rangle_h \Phi_*^{n+1,n})_j.$$

Substituting into the equation for ρ , we recovered

$$\delta_t^+ \rho_j^n - \delta_x [D_h d(\bar{\rho}_*^n) \delta_x \rho_*^{n+1} - \langle v_k \psi_1(v_k, \delta_x c_*^n) \rangle_h \Phi_*^{n+1,n}]_j = r_0 \rho_j^n \left(1 - \frac{\rho_j^n}{\rho_{\max}} \right)$$

the finite difference scheme for the volume-exclusion PKS model.

Positive preserving in the limit

Considering $q(\rho) = 1 - \rho/\bar{\rho}$ and $f(\rho) = r_0\rho \left(1 - \frac{\rho}{\rho_{\max}}\right)_+$

Proposition

- If $\rho_0^n \geq 0$ for all j , then, for whatever $\Delta t > 0$, we have

$$\rho_j^n \geq 0$$

for all j and $n \geq 1$.

- If we use a modified implicit scheme by changing $d(\bar{\rho}_*^n)$ to $d(\bar{\rho}_*^{n+1})$, where $d(\rho) = q(\rho) - \rho q'(\rho)$, it can be further proved that if $\rho_j^0 < \bar{\rho}$ for all j and $(1 + r_0\Delta t)\rho_{\max} < \bar{\rho}$, then

$$\rho_j^n \leq \bar{\rho}$$

for all j , and $n \geq 1$.

Generalization to 2D

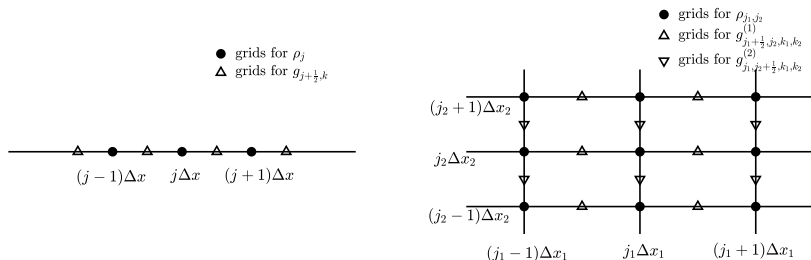


Figure: Grids in \mathbf{x} -space for computing $\rho(t, \mathbf{x})$ and $g(t, \mathbf{x}, \mathbf{v})$ in 1D and 2D.

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Numerical setup for 1D tests

- Domain $(x, v) \in (-20, 20)^2$ with a uniform mesh $\Delta x = 0.1, \Delta v = 0.2$
- Periodic boundary condition in x and the zero boundary condition in v
- Parameters: $r_0 = 0.1, \rho_{\max} = 0.5, \bar{\rho} = 1$ and

$$\psi_0(v) = \frac{1}{\sqrt{2\pi}} e^{-\frac{v^2}{2}}, \quad \psi_1(v, \nabla c) = \frac{v}{\sqrt{2\pi}} e^{-\frac{v^2}{2}} \nabla c.$$

- Initial data

$$\rho_0(x) = 0.5 + u(x), \quad c_0(x) = 0.5, \quad g_0^\varepsilon(x, v) = 0,$$

where $u(x) \in (-0.1, 0.1)$ is a uniformly distributed random function

Comparison of 1D solutions

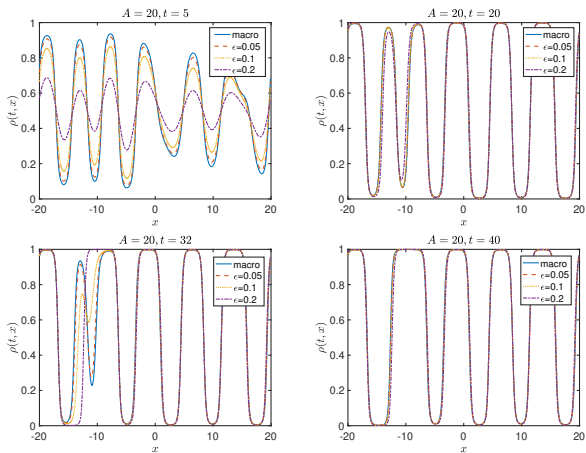


Figure: Comparison of ρ_{macro} and $\rho_{\text{kinetic}}^\epsilon$ for $A = 20$ and different ϵ 's. Here A is the chemotactic sensitivity.

Energy dissipation

Proposition

Define the energy

$$\mathcal{E}(t) = \int \Phi(\rho) \, dx - \frac{1}{2} \int \rho c \, dx ,$$

where

$$\Phi'(\rho) = H(\rho) := \frac{D_0}{\beta} \ln \left(\frac{\rho}{q(\rho)} \right) ,$$

we have

$$\frac{d}{dt} \mathcal{E}(t) = - \int \rho \chi(\rho) |\nabla(H - c)|^2 \, dx \leq 0 ,$$

when $r_0 = 0$ or $r_0 > 0$ but $\rho_{\max} + \rho_{\max}/\bar{\rho} \leq 1$

Numerical test of energy dissipation

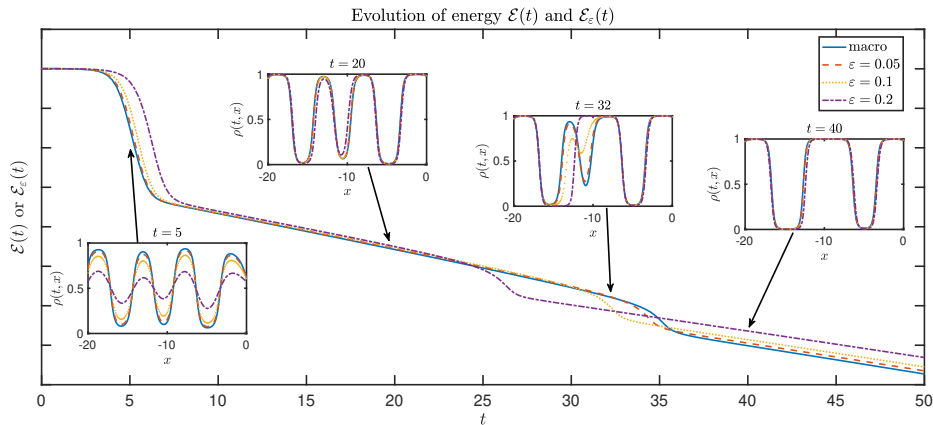


Figure: Evolution of $\mathcal{E}(t)$ and $\mathcal{E}_\varepsilon(t)$ along with the comparison between the kinetic solutions $\rho_{\text{kinetic}}^\varepsilon$ and the macroscopic solutions $\rho_{\text{macro}}(t, x)$ at $t = 5, 20, 32, 40$ for $A = 20$.

Convergence test in ε

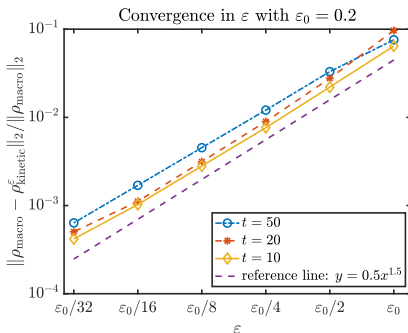
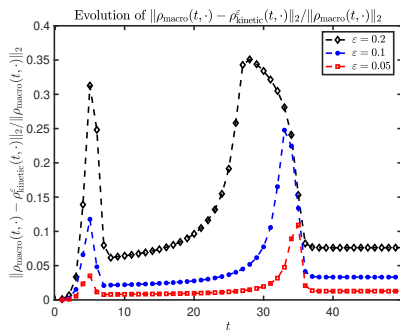


Figure: Left: Evolution of the relative L_2 -error $\frac{\|\rho_{\text{macro}}(t, \cdot) - \rho_{\text{kinetic}}(t, \cdot)\|_2}{\|\rho_{\text{macro}}(t, \cdot)\|_2}$ over time with $A = 20$. Right: Convergence of the relative L_2 -error in ε at $t = 10, 20, 50$.

Pattern formation from a perturbed initial data

- With a randomly perturbed initial data

$$\rho(t = 0, x) = 0.5 + u(x) ,$$

where $\rho \equiv 0.5$ is an equilibrium solution when $\rho_{\max} = 0.5$, a linear stability analysis ⁵ will give an analytical prediction of the pattern size.

- To numerically compute the pattern sizes, we consider using $1/k_{\max}$, where

$$k_{\max} = \operatorname{argmax}_{\lambda} (|\hat{\rho}(\lambda)|).$$

and $\hat{\rho}(\lambda) = \mathcal{F}(\rho)(t, x)$ is the Fourier transform of $\rho(t, x)$.

⁵L. Almeida, G. Estrada-Rodriguez, L. Oliver, D. Peurichard, A. Poulain, and F. Vallette (2020)

Numerical verification of pattern size

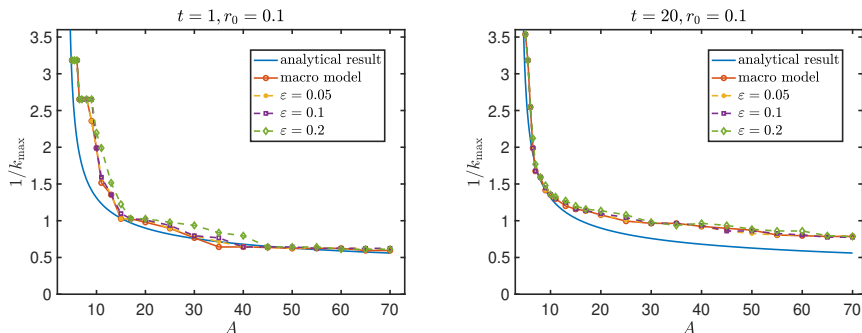


Figure: Comparison of $\frac{1}{k_{\max}}$ between the analytical, the numerical results of the macro and the kinetic model. The numerical result is computed by averaging 10 solutions, each evolved from a random initial data around 0.5. Here A is the chemotactic sensitivity.

Numerical setup for 2D tests

- Domain $\mathbf{x} \in (-20, 20)^2$, $\mathbf{v} \in (-10, 10)^2$ with a uniform mesh
- Periodic boundary condition in \mathbf{x} and the zero boundary condition in \mathbf{v}
- Parameters: $r_0 = 0.1$, $\rho_{\max} = 0.5$, $\bar{\rho} = 1$ and

$$\psi_0(\mathbf{v}) = \frac{1}{2\pi} e^{-\frac{|\mathbf{v}|^2}{2}}, \quad \psi_1(\mathbf{v}, \nabla_{\mathbf{x}} c) = \frac{\mathbf{v}}{2\pi} e^{-\frac{|\mathbf{v}|^2}{2}} \cdot \nabla_{\mathbf{x}} c.$$

- Initial data

$$\rho_0(\mathbf{x}) = 0.5 + u(\mathbf{x}), \quad c_0(\mathbf{x}) = 0.5, \quad g_0^\varepsilon(\mathbf{x}, \mathbf{v}) = 0,$$

where $u(\mathbf{x}) \in (-0.1, 0.1)$ is a uniformly distributed random function

Chemotactic sensitivity and patterns

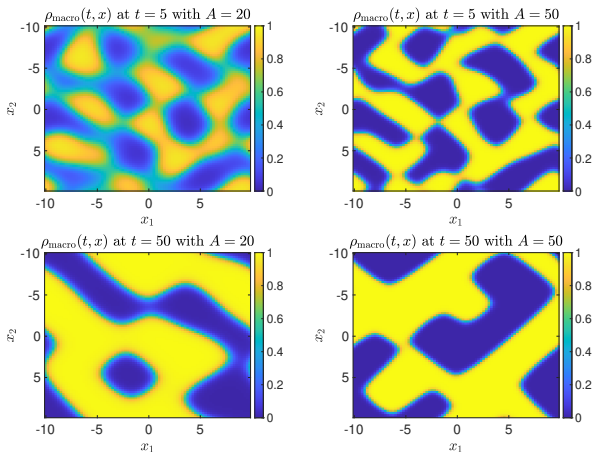


Figure: Plots of density $\rho_{\text{macro}}(t, x)$ at $t = 5, 50$ (from top to bottom) with $A = 20, 50$ (from left to right), respectively.

Patterns for different initial data

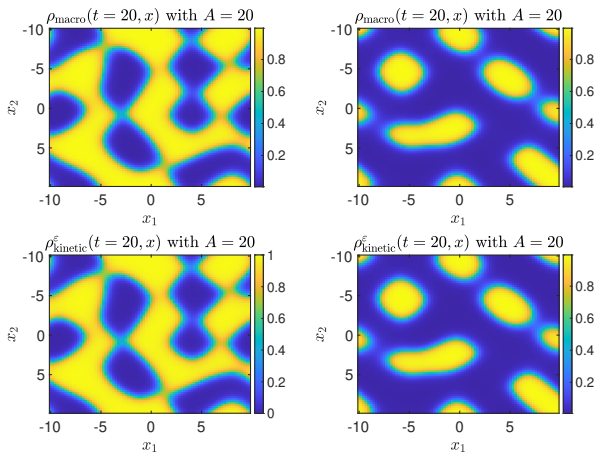


Figure: Comparison between $\rho_{\text{macro}}(t, \mathbf{x})$ (first row) and $\rho_{\text{kinetic}}^{\varepsilon}(t, \mathbf{x})$ with $\varepsilon = 10^{-2}$ (second row) for different initial conditions $\rho_0 \approx 0.5$ (first column) and $\rho_0 \approx 0.1$ (last column) and different values of the chemotactic sensitivity $A = 20$.

Convergence test of 2D model

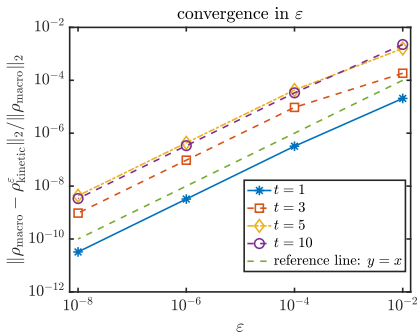
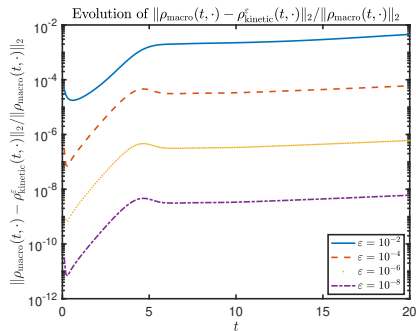


Figure: Left: Evolution of the relative L_2 -error. Right: Convergence of the relative L_2 -error in ε at $t = 1, 3, 5, 10$ with $A = 20$ and $\Delta t = 10^{-2}$.

Conclusions

- A kinetic model for chemotaxis incorporating a density dependence is derived.
- The volume-exclusion PKS model can be formally proved as the diffusion limit of the kinetic 'velocity-jump' model.
- An IMEX asymptotic preserving finite difference scheme based on a micro-macro decomposition is proposed for the kinetic model.
- In numerical experiments, we verified the convergence of the kinetic model, yet observed some differences, such as a delay in the aggregation process.

Future work:

- To extend the derivation to consider different turning kernels, to take into account cell-cell adhesion, etc.
- To construct numerical schemes for related cases.
- To connect with stochastic processes of interacting populations.

Reference: arXiv:2211.01415 [math.AP]

Thank you all for your attention!