A novel derivation of rigorous macroscopic limits from a micro-meso description of signal-triggered cell migration in fibrous environments

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Abstract

In this work we upscale a prototypical kinetic transport equation which models a cell population moving in a fibrous environment with a chemo- or haptotactic signal influencing both the direction and the magnitude of the cell velocity. The presented approach to scaling does not rely on orthogonality and treats parabolic and hyperbolic scalings in a unified manner. It is shown that the steps of the formal limit procedures are mirrored by rigorous operations with finite measures provided that the measure-valued position-direction fiber distribution enjoys some spacial continuity.

Keywords: cell movement, heterogeneous tissue, hyperbolic scaling, kinetic transport equations, measure-valued solutions, multiscale modelling; parabolic scaling, reaction-diffusion-taxis equations

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1 Introduction

1.1 Biological motivation and modelling aspects

Cell migration is a highly complex biological process involving a multitude of mechanisms and being influenced by manifold chemical and physical components of the extracellular environment. Cell motility is decisive for a plethora of physiological processes such as wound healing, regeneration of tissues, embryonic development, tumour growth and metastasis. Most of these involve interactions of the respective cells (e.g. fibroblasts, endothelial cells, chondrocytes, osteoblasts, tumour cells) with fibrous parts of the surrounding tissue, which they are able to modify in various ways, but also use as support and guidance for migration, proliferation, and even survival. Most cells adapt dynamically to changes in soluble (chemoattractants and/or -repellents) and insoluble (tissue) components of their environment, thus exhibiting a tactic behavior. Thereby they can perceive space-time variations of the respective signals, thus follow concentration/density gradients. Such motile behavior mediated by tissue and chemical signals is commonly termed haptotaxis and chemotaxis, respectively, and has been the object of numerous mathematical works concerned with modeling these phenomena and performing analysis and/or numerics for the obtained systems of differential equations. We refer, e.g. to [1, 22, 26, 29, 34] for reviews of taxis models from various perspectives.

One way to deduce such reaction-diffusion-taxis equations (RDTEs) is to consider a multiscale approach which starts from the microscopic scale of single cell behaviour (characterising cell trajectories and possibly also the so-called activity variables [2]) and obtains one or several (depending on the number of cell (sub)populations involved) corresponding kinetic transport equations (KTEs). Each of these equations is stated for a cell (sub)population distribution density depending on time, position, velocity, and the mentioned activity variables. A subsequent appropriate upscaling leads from this mesoscopic description to effective RDTEs on the macroscopic scale of cell population(s). Equations obtained in this manner inherit important low-level information from the original KTEs allowing for a more detailed and accurate modelling. Models for cell dispersal which were obtained in this framework have been proposed, e.g. in [5, 17, 21, 33, 36, 37, 40] for various scenarios of bacteria performing chemotaxis, some of these works also providing rigorous macroscopic limits. Still in the context of chemotactic behaviour, flux-limited macroscopic RDTEs have been obtained from KTEs e.g. in [3] by appropriate macroscopic

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limits; those systems have, among other advantages, the essential ability to enable finite propagation speed of the cell population performing the respective kind of diffusion and/or chemotaxis.

The dynamics of cell migration in fibrous tissues is more complex, as the cells perform several types of taxis, of which haptotaxis is essential [4]. Moreover, some cancer cells like glioma (generating the most common type of primary brain tumors) are able to exploit the underlying tissue anisotropy in order to enhance their motility [18], thus the orientation of tissue fibers is relevant and should be included in the modelling. First models employing KTEs and addressing this issue in the framework of cell migration were proposed in [6, 23, 25, 35]. Of these, in the former study a macroscopic RDTE involving chemo- and haptotaxis was obtained, whereas in the others some appropriate scaling led to RDTEs with myopic diffusion explicitly involving the space-dependent distribution of tissue fibers. Works [28, 31] extended the common KTE modelling framework to include both types of taxis (via adequate kernels) in the terms characterizing velocity reorientations. More recent multiscale models characterizing chemo- and/or haptotactic cancer cell migration in tissue networks and featuring myopic diffusion were proposed in [7–9, 13–15, 27, 30], all of which performed formal upscalings from the micro-meso- to the macroscopic.

In this work we deduce, first formally and then rigorously, macroscopic diffusion-taxis equations (DTEs) for cell migration in an anisotropic tissue. Following the multiscale modelling approach, we first construct a (mesoscopic) conservative linear KTE involving transport with respect to velocity and then perform parabolic (provided that the mean fiber orientation vanishes) and hyperbolic upscalings. The equations obtained in the macroscopic limit are accordingly diffusion- or drift-dominated. Recently, [9] shortly reviewed the hitherto available alternative ways to include cell level environmental influences in a KTE-based modelling framework leading to taxis terms in the macroscopic limit. Of those, the present note addresses the one which accounts for biochemical and/or biophysical effects translated into cell stress and forces acting on the cells. In the corresponding KTE this results in a non-zero transport term with respect to velocity. Such approach was also considered, e.g. in [6, 9, 11], the latter two actually combining it with the involvement of some further cell activity variables. Our model is closer to [11], yet accounts for less biological complexity. In particular, it does not include any activity variables. Nevertheless, cells considered here can actively change their velocity due to the presence of a stimulus. The aim of our work is to perform formal and rigorous upscalings for the model, presenting them in a transparent manner.

1.2 Mathematical aspects

1.2.1 Macroscopic approximations

Modelling with mesoscopic KTEs and subsequently upscaling them to get macroscopic RDTEs has proved to be a very effective scheme in the context of population migration in heterogeneous surroundings. However, the calculations are usually performed only formally and lacked rigorous justification. Overall, those settings which are very accurate from the modelling point of view are often particularly difficult to handle rigorously. One reason here is that since a rigorous limit procedure amounts to a proof of existence of solutions to the resulting macroscopic model, it is not surprising that it can be very difficult, if not impossible, to carry out if that model is nontrivial. An example of this is the study [39] of a one-dimensional haptotaxis system with a degenerate myopic diffusion arising from a potential tissue heterogeneity which was formally obtained through upscaling in [16]. In this model, the main challenges are a strong coupling of the equations and the strongly degenerate diffusion. Yet another illustrative example is a system describing particle motion in a fluid for which rigorous parabolic and hyperbolic scalings were conducted in [19] and [20], respectively. There the challenges on the macroscopic level are due to a strong coupling and the fact that one of the equations is an incompressible Navier-Stokes equation.

In this note we avoid the coupling issue and consider a single mesoscopic linear KTE for a single cell population moving in a heterogeneous external environment under the influence of an external signal. The biophysical/biochemical force in our KTE has a similar form to the drag Stokes force used in [19, 20]. It is proportional to the difference between the cell velocity and an external velocity field. In our case the field stems from a stimulus, not a fluid. Furthermore, unlike that model, not a diffusion with respect to velocity, but, rather, a turning operator is incorporated into the KTE. The latter describes the chaotic interactions with a heterogeneous extracellular environment, leading to a possibly degenerate myopic diffusion as in [25, 39]. To summarise, our equation can be categorised as a Vlasov-Fokker-Planck equation into which a linear turning operator has been incorporated. From the mathematical perspective, it can also be viewed as a variant of the linear Boltzmann-Maxwell equation, a linear kinetic equation with an external field. Such an equation describes a gas of charged particles moving under the influence of an external field through an unchanging background of another type of particles. However, it should be stressed that the interactions between cells and, e.g. tissue fibers are very different from collisions
between two kinds of physical particles. In particular, the directional distribution of the fibers is by no means a Maxwellian and can actually be very unsmooth. In this work we allow it to be a finite measure, such as, e.g., a Dirac delta function.

Establishing convergence of a sequence of linear KTEs such as ours to, e.g., a parabolic RDE can be complicated, as these equations are of a very different nature. Indeed, a KTE is an integro-differential equation that includes both a divergence operator, typically with respect to the time and space variables (in the transport part), and integration (in the turning operator) with respect to the velocity variable which needs to be scaled out. In our case due to the external forces the divergence with respect to the velocity variable is also present in the transport. This makes the upsampling even more challenging.

Currently, the most commonly applied upsampling procedures in the context of modelling cell movement are: the parabolic scaling, the hyperbolic scaling, and the moment closure method. Since our approach relies on rescaling, we do not consider the latter method and concentrate on the first two instead. As usual for such scalings, a small scaling parameter \( \varepsilon \) is introduced, the KTE rescaled, and the aim is then to approximate the solutions as \( \varepsilon \) is sent to zero. The approximations are typically sought in the form of Hilbert or Chapman-Enskog expansions. The traditional approach to both parabolic and hyperbolic limits relies on some Hilbert space structure already for the formal scalings, see e.g. the derivations presented in [25, 33]. In particular, the Chapman-Enskog expansion that is typically used in the hyperbolic case is based on the assumption that the actual limit function, i.e. the zero order term in the expansion, is orthogonal to the first order correction. However, this assumption cannot be justified in such models as the one we consider here. Our calculations show that the resulting leading zero order term and the first order correction are not mutually orthogonal in general. The first order correction has a rather complicated form, carrying the dependence on the gradient of an external macroscopic quantity.

A technique that has been extensively used when dealing with rigorous scalings of linear KTEs is the semigroup theory [32]. However, the presence of the derivative with respect to the velocity variable turns out to be problematic already in very basic cases, such as the hyperbolically scaled one-dimensional linear KTE (10.1.9) from [32, Chapter 10]. Unlike our model, there the velocity field is taken to be constant, so that it does not depend on the velocity variable, and a regular Maxwellian distribution is used in the collision term. Already in this simplified situation it turns out necessary to work in the phase space which is a weighted \( L^1 \) space in order to pass rigorously to a macroscopic limit. In our case, we are forced to work in even rougher spaces of Radon measures.

In this note we present a novel and unified approach towards both parabolic and hyperbolic scalings, which avoids the necessity to work in Hilbert spaces and does not use semigroups. Its key component is a differential equation connecting moments of order zero and two of the KTE solutions, which at first glance resembles a damped wave equation with a transport term. Differentiating this equation a sufficient number of times with respect to the scaling parameter and passing to the limit allows to obtain the coefficients of the Hilbert expansion of any order. It turns out that thanks to the linearity of the KTE and the mass preservation, under rather general assumptions the derivation steps can then be mirrored by operations with finite measures leading to a rigorous scaling. Our work extends a result obtained in [24] where a parabolic scaling was performed in the special case of a (non-evolving) space-homogeneous fiber distribution and for a KTE without velocity derivative.

### 1.2.2 Mesoscopic approximation

Often, when considering e.g., Hilbert expansions of the form \( \psi = \psi^0 + \varepsilon \psi^1 + O(\varepsilon^2) \) for the solution of the original KTE, one would like to combine the leading order term \( \psi^0 \) and first order correction \( \varepsilon \psi^1 \) in order to get a mesoscopic approximation. The basic straightforward approach which consists of directly adding them together is known to have a serious drawback: the resulting function is not necessarily nonnegative. One way to avoid this issue is to use a nonlinear Hilbert expansion as was done, e.g. in [10] for a simpler equation and in space dimension one. In this work we develop an alternative approach which requires dealing with a truncated version of the original mesoscopic equation. This new PDE is a transport equation in terms of a mesoscopic first order approximation. It preserves both positivity and the total mass and can be solved numerically after the macroscopic approximations of zero and first orders are obtained.

The remainder of the paper is structured as follows. To begin with, we introduce in Section 2 a prototypical KTE describing cell movement on the mesolevel. For this model, we consider in Section 3 parabolic and hyperbolic scalings of time and space and formally derive the corresponding limits and first-order corrections, also developing a mesoscopic first order approximation which preserves both positivity and the total mass. The rigorous scaling is done in Section 4. In the closing Section 5 we discuss our findings.
2 A KTE modeling framework

In this Section we describe the KTE modelling framework which is the starting point for our approach. More precisely, we consider the migration of cancer cells in a fibrous, anisotropic tissue under the influence of some extracellular signal. The description of single cell behaviour involves the position and velocity dynamics which allow to reconstruct cell trajectories. Without loss of generality we assume the cell speed magnitude to be less than one and consider the velocity space

\[ V = B_1(0) := \{ \mathbf{v} = s \mathbf{\theta} : \ s \in [0, 1), \ \mathbf{\theta} \in S_1(0) \}, \quad S_1(0) := \{ \theta \in \mathbb{R}^n : |\theta| = 1 \}. \]  

(2.1)

For any quantity \( u \) depending on \( v \) we will denote

\[ \overline{u} := \int_{B_1(0)} u(v) \, dv. \]  

(2.2)

In order to capture with some accuracy the evolution of a large cell population, we start our modelling on the microscopic scale. Disregarding for the moment the chaotic interactions with tissue, we assume the movement of each cell to be described by the following system of ordinary differential equations corresponding to Newton’s second law:

\[
\frac{dx}{dt} = v, \\
\frac{dv}{dt} = S(t, x, v),
\]

(2.3a)

(2.3b)

where

\[ S(t, x, v) = -a(v - v_a(t, x)), \quad a > 0, \ v_a(t, x) \in B_1(0) \text{ for all } t > 0, \ x \in \mathbb{R}^n. \]  

(2.4)

Here \( t, x, v \), and \( S \) are time, position in space, velocity, and acceleration of a cell, respectively. Our choice of \( S \) is similar to the one in [11]. It reflects the cell tendency to redirect its velocity in order to realign with a certain preferred direction \( v_a \) and to decelerate if such a direction is missing. Most of the previous constructions assumed zero acceleration or such that vanishes at least in the radial direction, like, e.g., in [9], or does not depend on \( v \), like in [6]. For (2.4) this is clearly not the case, so that in our model a cell can change not only the direction, but also its speed and the change depends on the velocity itself. As we will see later in Section 3, this makes calculations more involved.

As in [6, 11], we model \( v_a \) in such a way as to account for the effect of chemo- or haptotaxis. The model proposed in [6] considers for the cell velocity dynamics a ‘chemotactic force’ depending on the gradient of some given chemical profile, hence also allowing for changes in the cell speed. The setting in [11] combines such ‘tactic forces’ (repellent chemotaxis and haptotaxis) with a repellent force caused by large gradients of macroscopic cell density, all these influences contributing to modifications of direction as well as speed of the migrating cells.

Aiming to describe the effect of the (macroscopic) tissue density or, alternatively, of some (possibly different) signal on the cell velocity reorientation, here we choose \( v_a \) in the form

\[ v_a = \mathbf{F} \frac{\nabla_v Q}{1 + |\nabla_v Q|}, \]

(2.5)

where \( Q = Q(t, x) \) is the concentration of a chemoattractant or the macroscopic density of tissue (one could also consider both, introducing yet another macroscopic quantity, \( A(t, x) \) for that purpose) and \( \mathbf{F} = \mathbf{F}(x) \) is a tensor such that

\[ \|
\mathbf{F}(x)\|_2 \leq 1 \quad \text{for all } x \in \mathbb{R}^n. \]

As in some previous works [9, 11], the tensor \( \mathbf{F} \) is supposed to model, e.g., biomechanical cell stress or simply have a deviatoric effect due to the heterogeneity of the environment. For instance, when modelling the glioma spread, one could choose (similarly to [11]), for each brain voxel centered at \( x \), the matrix \( \mathbf{F}(x) \) to be the water diffusion tensor obtained by diffusion tensor imaging (DTI), a variant of MRI which is a standard noninvasive diagnostic tool. There is abundant evidence [18] that migrating glioma cells follow the highly anisotropic brain structure, particularly that of white matter. One could also choose the tensor \( \mathbf{F} \) to be velocity-dependent, e.g., taking

\[ \mathbf{F} := \frac{1}{a} \left( |v|^2 \mathbf{I}_n - v \otimes v \right), \]
as it was done in [9]. Here and in what follows $I_n$ denotes the identity matrix and $\otimes$ stands for the tensor product. We refer to that work for further details. In any case, the eigenvalues and eigenvectors of $F$ are used to encode relevant information about local anisotropy and diffusivity, thus helping to reconstruct the whole structure of brain tissue and opening the way for predictions about the extent of tumour spread. For more details on this approach we refer to [9, 13, 14, 35].

We model the fibrous extracellular environment by way of the orientational distribution of tissue fibers, upon taking

$$q = q(x, \hat{v}), \quad \dot{q} = \frac{v}{|v|},$$

$$q \geq 0, \quad \int_{S_1(0)} q(\theta) \, d\theta = 1,$$  

hence

$$\int_{B_1(0)} q(v) \, dv = \frac{1}{n}.$$  

For later purposes we introduce the moments

$$E[q] := \int_{S_1(0)} \theta q(\theta) \, d\theta,$$  

$$D[q] := \int_{S_1(0)} \theta \otimes \theta q(\theta) \, d\theta,$$  

$$T[q] := \int_{S_1(0)} \theta \otimes \theta \otimes \theta q(\theta) \, d\theta.$$  

Here $E[q]$ can be interpreted as the average orientation of tissue fibers, while $D[q]$ is the auto-correlation matrix and thus related to the variance-covariance matrix

$$V[q] := \int_{S_1(0)} (\theta - E[q]) \otimes (\theta - E[q]) q(\theta) \, d\theta,$$

so that

$$D[q] = V[q] + E[q] \otimes E[q].$$

As usual, when performing a parabolic scaling, we assume that

$$E[q] = 0,$$

which implies that

$$V[q] = D[q].$$

Finally, if $E[q] = 0$, then the third moment, $T[q]$ accounts for the skewness of distribution $q$, i.e. for its asymmetry about its mean.

Now we have all ingredients necessary to construct a model on the mesoscale. On this level we model the probability density function $c = c(t, x, v)$ for cell position and velocity over time. Our equation for $c$ reads:

$$\nabla_{(t,x,v)} \cdot ((1, v, S(t,x,v)) c) = \hat{c}_t c + \nabla_x \cdot (vc) - a \nabla_v \cdot ((v - v_*) c) = nqF - c.$$  

The left-hand side of (2.13) describes the mass-conserving transport along the trajectories generated by (2.3). The right-hand side captures chaotic interactions with tissue fibers. This is done by means of incorporating a turning operator with a turning kernel $nq$ and a constant turning rate (which for simplicity we set to 1). Thus, as in many previous works (see e.g. [7–9, 13–15, 23, 27, 30, 35]) we assume that the cells adapt their respective direction of motion to the local orientation of tissue fibers. Moreover, still in line with previous works, we consider the cell population to be compactly supported on the velocity space, thus introduce the boundary condition

$$c = 0 \quad \text{for } v \in \partial B_1.$$  

(2.14)
3 Formal upscaling

In this section we derive a formal macroscopic limit for the suitably rescaled KTE \((3.13)\). A rigorous argument requires adequate assumptions on the model parameters and is based on calculations with Radon measures. We postpone this issue to Section 4.

In order to set the frame we introduce some macroscopic time and space scales: for \(\varepsilon \in (0, 1)\) let
\[
\hat{t} = \varepsilon^\kappa t, \quad \kappa \in \{1, 2\},
\]
\[
\hat{x} = \varepsilon x.
\]
Rescaling \((3.13)\) and \((2.5)\) and dropping the hats leads to
\[
\varepsilon^\kappa \partial_t \varepsilon^\kappa + \varepsilon \nabla \cdot (v \varepsilon^\kappa) - a \nabla v \cdot ((v - \vec{v}) \varepsilon) = n q \varepsilon - \varepsilon^\kappa.
\]
where
\[
v^\kappa = \frac{\varepsilon \nabla_\kappa Q}{1 + \varepsilon |\nabla_\kappa Q|} = \varepsilon \nabla_\kappa Q - \varepsilon^2 \nabla_\kappa Q |\nabla_\kappa Q| + O (\varepsilon^3).
\]

3.1 Equations involving moments

To begin with, we integrate \((3.3)\) by parts with respect to \(v\) over \(B_1(0)\) and divide by \(\varepsilon\) in order to obtain an equation which connects the moments of orders zero and one:
\[
\varepsilon^\kappa \partial_t \varepsilon^\kappa + \varepsilon \nabla \cdot (v \varepsilon^\kappa) = 0.
\]
Thereby we used the fact that \(\varepsilon\) vanishes on the boundary. Next, we multiply \((3.3)\) by \(v\) and once again integrate by parts over \(B_1(0)\):
\[
\varepsilon^\kappa \partial_t \varepsilon^\kappa + \varepsilon \nabla \cdot (v^T \varepsilon^\kappa) = \varepsilon^2 \nabla Q |\nabla Q| + O (\varepsilon^3).
\]
Rearranging and dividing \((3.6)\) by \(\varepsilon^{\kappa-1}\) leads to
\[
-(a + 1) \varepsilon^{\kappa-1} \nabla \cdot (v^T \varepsilon) = \varepsilon^{\kappa} \nabla \cdot (v^T \varepsilon) = \varepsilon^{\kappa} \varepsilon^\kappa (a v^\kappa + \frac{n}{n+1} E[v]) \varepsilon - \varepsilon^{\kappa} \nabla \cdot (a v^\kappa + \frac{n}{n+1} E[v]) \varepsilon.
\]
Next, we apply \((\nabla \cdot v)\) to both sides of \((3.7)\) and plug the expression on the right-hand side into \((3.5)\). Thus we arrive at the following differential equation for the moments of zero and second order:
\[
\varepsilon^\kappa \partial_t \varepsilon^\kappa + (a + 1) \partial_t \varepsilon^\kappa = \varepsilon^2 \nabla \cdot (v^T \varepsilon) - \varepsilon^{\kappa-1} \nabla \cdot \varepsilon^\kappa (a v^\kappa + \frac{n}{n+1} E[v] q) \varepsilon.
\]
Remark 3.1. At first glance, equation \((3.8)\) is a damped wave equation with a transport term. Yet this is only the case if \(\kappa = 1\) and \(\varepsilon\) is independent of \(v\), so that the first term on the right-hand side becomes \(\varepsilon^2 \Delta \varepsilon \).

3.2 Zero order approximation

Passing formally to the limit as \(\varepsilon \to 0\) in \((3.3)\) and using \((3.4)\) we obtain that
\[
c^0 := \lim_{\varepsilon \to 0} c^\varepsilon
\]
satisfies the equation
\[
-a \nabla v \cdot (v c^0) = n q c^0 - c^0,
\]
which can be solved explicitly:

**Proposition 3.2.** For any fixed \(c^0\) there exists a unique solution to \((3.9)\) with \(c^0 = 0\) for \(|v| = 1\):
\[
c^0 = c^0 q \xi_1,
\]
where
\[
\xi_1(v) = \begin{cases} \frac{n}{n-\alpha} \left(\frac{|v|}{\alpha} \right)^{-\alpha+\frac{1}{\alpha}} - 1 & \text{for } \alpha \neq \frac{1}{n}, \\ -\frac{n}{\alpha} \ln |v| & \text{for } \alpha = \frac{1}{n}. \end{cases}
\]
Proof. We use the method of characteristics which leads to the following ODE system:
\begin{align}
\dot{v} &= -av, \quad |v(0)| = 1, \tag{3.12a} \\
\dot{c} &= nc^\sigma + (na - 1)c, \quad c(0) = 0. \tag{3.12b}
\end{align}

The ODE (3.12a) leads to
\[ v(\sigma) = e^{-a\sigma} v(0), \tag{3.13} \]
so that
\[ \left| v(\sigma) \right| = e^{-a\sigma}. \tag{3.14} \]

Let \( a \neq \frac{1}{n} \). Solving the ODE (3.12b), we obtain using (2.6) and (3.13) that
\begin{align*}
\dot{c}(\sigma) &= \frac{1}{\alpha} \int_0^\sigma ne^{(na-1)(\sigma-\tau)} q(\tilde{v})(\tau) d\tau \\
&= \frac{1}{\alpha} q(\tilde{v})(\sigma) \int_0^\sigma ne^{(na-1)(\sigma-\tau)} d\tau \\
&= \frac{1}{\alpha} q(\tilde{v})(\sigma) \frac{n}{na-1} \left( e^{(na-1)\sigma} - 1 \right), \tag{3.15}
\end{align*}
which together with (3.14) gives (3.10) for \( a \neq \frac{1}{n} \). Passing to the limit as \( a \to \frac{1}{n} \) gives the formula for \( a = \frac{1}{n} \). \( \square \)

Next, we multiply (3.9) by \( v \) and \( vv^T \), respectively, and integrate by parts over \( B_1(0) \) in order to find the moments of order one:
\begin{align*}
\frac{1}{n} \dot{v} c &= \frac{1}{n+1} \mathbb{E}[q] c^\sigma - vc^\sigma \\
\Leftrightarrow \quad \frac{1}{\alpha} \dot{v} c &= \frac{1}{\alpha + 1} \frac{n}{n+1} \mathbb{E}[q] c^\sigma, \tag{3.16}
\end{align*}
and two, respectively:
\begin{align*}
\frac{2}{n} \dot{v} v c &= \frac{1}{n+2} \frac{n}{n+1} \mathbb{E}[q] c^\sigma - vv^T c^\sigma \\
\Leftrightarrow \quad \frac{2}{n+2} \dot{v} v c &= \frac{1}{n+2} \frac{n}{n+1} \mathbb{E}[q] c^\sigma. \tag{3.17}
\end{align*}

Passing formally to the limit in (3.8), using (3.4) and (3.17) we arrive at a drift-diffusion equation
\[(a+1)\dot{c} c = \frac{1}{2a+1} \frac{n}{n+2} \nabla_x \nabla_x^T : (\mathbb{D}[q] c^\sigma) - a \nabla_x \cdot (c^\sigma \nabla_x Q) \quad \text{if } \kappa = 2 \quad \text{and } \mathbb{E}[q] = 0. \tag{3.18}\]

Passing formally to the limit in (3.5) and using (3.16) we arrive at a drift equation
\[(a+1)\dot{c} c + \frac{n}{n+1} \nabla_x \cdot (c^\sigma \mathbb{E}[q]) = 0 \quad \text{if } \kappa = 1. \tag{3.19}\]

Several remarks on the resulting equations are in order.

Remark 3.3. In the parabolic scaling case (\( \kappa = 2 \)) the first term on the right-hand side of equation (3.18) represents the myopic diffusion which, if the scaling constant depending on \( n \) and \( a \) is neglected, can be decomposed as follows:
\[ \nabla_x \nabla_x^T : (\mathbb{D}[q] c^\sigma) = \nabla_x \cdot (\mathbb{D}[q] \nabla_x c^\sigma) + \nabla_x \cdot (c^\sigma \nabla_x : \mathbb{D}[q]), \]
where the first summand is the regular anisotropic diffusion in divergence form with diffusion coefficient \( \mathbb{D}[q] \), and the second summand describes cell transport with velocity \( \nabla_x \cdot c^\sigma \mathbb{E}[q] \). Due to (2.12) we have that \( \mathbb{D}[q] \) coincides with the variance-covariance matrix \( \mathbb{V}[q] \). Both effects are thus dependent on the orientation distribution of tissue fibers. The second term on the right-hand side of (3.18) describes the taxis towards the gradient of the macroscopic quantity \( Q \) (e.g., haptotaxis if \( Q \) represents macroscopic tissue density, chemotaxis if it denotes the concentration of some chemoattractant). When a hyperbolic scaling (\( \kappa = 1 \)) is performed, however, the dynamics is dominated by transport in the mean fiber direction - at least at leading order. In the next Subsection 3.3 we will derive first order corrections for both scalings, which allows for more accurate approximations.
Remark 3.4 (The role of $a$). Both scaling limits involve a constant $a$, which is a scaling parameter for the acceleration. Specifically, $1/a$ can be seen to be analogous to the parameter employed in $[9]$ to characterise single cell velocity dynamics: it should be a quantity of the order $1/c\gamma$, for some $\gamma > 0$ representing a constant related to smaller scales, e.g. microtubule extension zones that are responsible for the subcellular level exchange of cells with their environment. For further details we refer to $[9]$. For both types of scaling, sending $a$ to zero leads to the standard diffusion and drift equations, which were previously derived for cell movement without transport with respect to velocity included in the KTE, see e.g. $[25]$. Conversely, for large $a$ the role of terms depending on the (mesoscopic) fiber orientation becomes negligible. In the parabolic limit the taxis with respect to the macroscopic quantity $Q$ then fully dominates the space-time evolution. In the hyperbolic scaling case no such additional effect is present, so that the macroscopic cell density remains nearly constant over time. 

Remark 3.5 (Directed/un-directed fibers). As in previous studies, see, e.g. $[23]$, the parabolic scaling can only be performed under the assumption $E[q] = 0$. This occurs, but not exclusively, in the case where the tissue fibers are un-directed. By this we mean (as, e.g. in $[23]$) that the fibers are symmetrical all along their axes, i.e. there is no ‘up’ and ‘down’ on such fibers, which translates into symmetry of the orientational distribution:

$$q(x, \theta) = q(x, -\theta) \quad \text{for all } \theta \in S_1(0).$$

On the other hand, the hyperbolic scaling required no such assumption.

As described in $[25]$, the choice of an appropriate scaling can be made based on measurements of reference values (such as speeds, turning rates, etc.) in a specific application. As far as brain tissue is concerned, it is still not clearly established whether it is directed or not, however recent mathematical modelling and simulations of typical glioblastoma patterns suggests that it might be un-directed $[30]$. 

### 3.3 First order correction

The above formal passage to the limit for $\varepsilon \to 0$ has led to macroscopic PDEs only containing leading order terms. To obtain first order corrections, hence enhanced approximations, we start by introducing

$$c_1^0 := \lim_{\varepsilon \to 0} \varepsilon c^\varepsilon,$$

$$c_{01} := c^0 + \varepsilon c_1^0.$$

Differentiating (3.3) and (3.8) with respect to $\varepsilon$, letting $\varepsilon \to 0$, and using (3.4) we obtain: from (3.3) that

$$-a\nabla v \cdot (v c_1^0) = na c_1^0 - \delta_{1\kappa} \partial_t c_0^0 + \nabla_x \cdot (v c_0^0) + aF \nabla_x Q \cdot \nabla_v c_0^0,$$

(3.20)

where $\delta_{1\kappa}$ denotes the Kronecker delta, and from (3.8) that

$$(a + 1)\partial_t c_1^0 = \nabla_x \nabla^T v : \nabla^T v c_1^0 - a \nabla_x \cdot \left( \nabla^T v Q \nabla_x Q \right) + a \nabla_x \cdot \left( \nabla^T v Q \nabla_x Q \right)$$

if $\kappa = 2$ and $E[q] = 0$, 

(3.21)

$$(a + 1)\partial_t c_1^0 + (a + 1)\partial_t c_1^0 = \nabla_x \nabla^T v : \nabla^T v c_1^0 - a \nabla_x \cdot \left( \nabla^T v Q \nabla_x Q \right)$$

if $\kappa = 1$. 

(3.22)

Proposition 3.6. For any $c_1^0$ there exists a unique solution to (3.20) with $c_1^0 = 0$ for $|v| = 1$:

$$c_1^0 = c_1^0 q_1 - \delta_{1\kappa} \partial_t c_0^0 q_2 - \nabla_x \cdot \left( v q_0 \right) \xi_3 - aF \nabla_x Q \cdot \nabla_v q_4 + v q_5,$$

(3.23)

where

$$\xi_2(v) = \begin{cases} \frac{n}{n+1} |v|^{-n+\frac{1}{2}} \left( -\ln |v| + \frac{1}{n+1} \left( |v|^{-n+\frac{1}{2}} - 1 \right) \right) & \text{for } a \neq \frac{1}{n}, \\ \frac{n}{2} \ln^2 |v| & \text{for } a = \frac{1}{n}, \end{cases}$$

(3.24)

$$\xi_3(v) = \begin{cases} \frac{n}{n+1} |v|^{-n+\frac{1}{2}+1} \left( 1 - |v| + \frac{1}{n+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 \right) \right) & \text{for } a \neq \frac{1}{n+1}, \\ \frac{n}{n+1} |v|^{-n+\frac{1}{2}+1} \left( 1 - |v| + \frac{1}{n+1} |v| \ln |v| \right) & \text{for } a = \frac{1}{n+1}, \end{cases}$$

(3.25)

$$\xi_4(v) = \begin{cases} \frac{n}{n+1} |v|^{-n+\frac{1}{2}+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 + \frac{1}{n+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 \right) \right) & \text{for } a \neq \frac{1}{n+1}, \\ \frac{n}{n+1} |v|^{-n+\frac{1}{2}+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 + \frac{1}{n+1} |v| \ln |v| \right) & \text{for } a = \frac{1}{n+1}, \end{cases}$$

(3.26)

$$\xi_5(v) = \frac{n}{n+1} |v|^{-n+\frac{1}{2}+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 + \frac{1}{n+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 \right) \right) \text{ for } a \neq \frac{1}{n+1},$$

(3.27)

for some

$$\xi_1(v) = \begin{cases} \frac{n}{n+1} |v|^{-n+\frac{1}{2}} \left( -\ln |v| + \frac{1}{n+1} \left( |v|^{-n+\frac{1}{2}} - 1 \right) \right) & \text{for } a \neq \frac{1}{n}, \\ \frac{n}{2} \ln^2 |v| & \text{for } a = \frac{1}{n}, \end{cases}$$

(3.24)

$$\xi_3(v) = \begin{cases} \frac{n}{n+1} |v|^{-n+\frac{1}{2}+1} \left( 1 - |v| + \frac{1}{n+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 \right) \right) & \text{for } a \neq \frac{1}{n+1}, \\ \frac{n}{n+1} |v|^{-n+\frac{1}{2}+1} \left( 1 - |v| + \frac{1}{n+1} |v| \ln |v| \right) & \text{for } a = \frac{1}{n+1}, \end{cases}$$

(3.25)

$$\xi_4(v) = \begin{cases} \frac{n}{n+1} |v|^{-n+\frac{1}{2}+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 + \frac{1}{n+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 \right) \right) & \text{for } a \neq \frac{1}{n+1}, \\ \frac{n}{n+1} |v|^{-n+\frac{1}{2}+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 + \frac{1}{n+1} |v| \ln |v| \right) & \text{for } a = \frac{1}{n+1}, \end{cases}$$

(3.26)

$$\xi_5(v) = \frac{n}{n+1} |v|^{-n+\frac{1}{2}+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 + \frac{1}{n+1} \left( |v|^{-n+\frac{1}{2}+1} - 1 \right) \right) \text{ for } a \neq \frac{1}{n+1},$$

(3.27)
\textbf{Remark 3.7.} Notice that the first order correction $c^1$ and leading term $c^0$ given by (3.23) and (3.10), respectively, are clearly not mutually orthogonal with respect to the scalar product of the weighted $L^2$-space $L^2(V; \frac{dx}{n\gamma}).$

\textit{Proof.} (of Proposition 3.6) Observe first that (2.6) implies
\[ \nabla_v q(\hat{v}) = (\hat{u}_\lambda - \hat{v}^T) \nabla_v q(\hat{v}) |v|^{-1}. \] (3.28)

Hence
\[ \nabla_v (q(\hat{v}) \xi_1(|v|)) = (\hat{u}_\lambda - \hat{v}^T) \nabla_v q(\hat{v}) |v|^{-1} \xi_1(|v|) + q(\hat{v}) \hat{c}_1^1(|v|). \] (3.29)

Using the method of characteristics and (3.10), (3.13), and (3.29) we obtain that
\[ c_1^0(\sigma) = \int_0^\sigma e^{(n\alpha-1)(\sigma-\tau)} \left( nq \varphi_n \hat{c}_1^1 - (\delta_{1\alpha} \hat{c}_1^0 \nabla_x \cdot (v \sigma^0) + a\overline{\nabla}_x Q \cdot \nabla_v \sigma^0) \right) (\tau) d\tau \]
\[ = \hat{c}_1^0(\xi_1) - \int_0^\sigma e^{(n\alpha-1)(\sigma-\tau)} \left( (\delta_{1\alpha} \hat{c}_1^0 \nabla_x \cdot (v \sigma^0) + a\overline{\nabla}_x Q \cdot \nabla_v \sigma^0) \right) d\tau \]
\[ = \hat{c}_1^0(\xi_1) - \delta_{1\alpha} \hat{c}_1^0 \nabla_v \xi_2 - \nabla_x \cdot (v \sigma^0) \xi_3 - a\overline{\nabla}_x Q \cdot (\nabla_v \xi_4 + v \xi_5), \] (3.30)

where
\[ \xi_2(v)(\sigma) = e^{(n\alpha-1)\sigma} \int_0^\sigma e^{-(n\alpha-1)\tau} \xi_1(|v|)(\tau) d\tau, \]
(3.31)
\[ \xi_3(v)(\sigma) = |v|^{-\frac{1}{2}} e^{(n\alpha-1)\sigma} \int_0^\sigma e^{-(n\alpha-1)\tau} |v| \xi_1(|v|)(\tau) d\tau, \]
(3.32)
\[ \xi_4(v)(\sigma) = |v| \sigma e^{(n\alpha-1)\sigma} \int_0^\sigma e^{-(n\alpha-1)\tau} |v|^{-1} \xi_1(|v|)(\tau) d\tau, \]
(3.33)
\[ \xi_5(v)(\sigma) = e^{(n\alpha-1)\sigma} \int_0^\sigma e^{-(n\alpha-1)\tau} \xi_1(|v|)(\tau) d\tau. \]
(3.34)

Since $e^{-(n\alpha-1)\tau} = |v|^{-\frac{1}{2}}(\tau)$ and $e^{-(n\alpha-1)\tau} d\tau = -\frac{1}{3}|v|^{-\frac{3}{2}} I d|v|$ due to (3.14), we can rewrite (3.31)-(3.34) as follows:
\[ \xi_2(v) = \frac{1}{a} |v|^{-\frac{n+1}{4}} \int_{|v|}^{\frac{1}{a}} s^{-\frac{n+1}{4}} \xi_1(s) ds, \]
(3.35)
\[ \xi_3(v) = \frac{1}{a} |v|^{-\frac{n+1}{4}} \int_{|v|}^{\frac{1}{a}} s^{-\frac{n+1}{4}} \xi_1(s) ds, \]
(3.36)
\[ \xi_4(v) = \frac{1}{a} |v|^{-\frac{n+1}{4}} \int_{|v|}^{\frac{1}{a}} s^{-\frac{n+1}{4}} \xi_1(s) ds, \]
(3.37)
\[ \xi_5(v) = \frac{1}{a} |v|^{-\frac{n+1}{4}} \int_{|v|}^{\frac{1}{a}} s^{-\frac{n+1}{4}} \xi_1(s) ds. \]
(3.38)

Let $a \neq \frac{1}{n}$. Plugging (3.11) into (3.35)-(3.38), we obtain
\[ \xi_2(v) = \frac{1}{a n a - 1} |v|^{-\frac{n+1}{4}} \int_{|v|}^{\frac{1}{a}} s^{-\frac{n+1}{4}} \left( s^{-\frac{n+1}{4}} - 1 \right) ds, \]
(3.39)
\[ \xi_3(v) = \frac{1}{a n a - 1} |v|^{-\frac{n+1}{4}} \int_{|v|}^{\frac{1}{a}} s^{-\frac{n+1}{4}} \left( s^{-\frac{n+1}{4}} - 1 \right) ds, \]
(3.40)
\[ \xi_4(v) = \frac{1}{a n a - 1} |v|^{-\frac{n+1}{4}} \int_{|v|}^{\frac{1}{a}} s^{-\frac{n+1}{4}} \left( s^{-\frac{n+1}{4}} - 1 \right) ds, \]
(3.41)
\[ \xi_5(v) = -\frac{1}{a^2} |v|^{-\frac{n+1}{4}} \int_{|v|}^{\frac{1}{a}} s^{-2} ds. \]
(3.42)

Finally, computing the integrals in (3.35)-(3.38), we arrive at (3.24)-(3.27) for $a \neq \frac{1}{n}$. Passing to the limit as $a \to \frac{1}{n}$ gives the formulas for $a = \frac{1}{n}$. \qed
Next, we multiply (3.20) by \( v \) and \( vv^T \), respectively, integrate by parts over \( B_1(0) \), and use (3.16) and (3.17) in order to find the moments of order one:

\[
\begin{align*}
\frac{\partial vv^T c_1}{\partial t} = & \frac{n}{n+1} E[q] c_1^T - \frac{n+1}{n+1} \left( \delta_1 \hat{c}_1 v c_1^T + \nabla_x \cdot vv^T c_1^T - ac^T \nabla_x Q \right) \\
\Rightarrow (a+1)vv^T c_1 = & \frac{n}{n+1} E[q] c_1^T - \left( \delta_1 \hat{c}_1 v c_1^T + \nabla_x \cdot vv^T c_1^T - a \left( \tilde{\nabla}_x Q c_1^T + \nabla_x Q \nabla_x Q^T \right) \right) \\
\end{align*}
\]

and two, respectively:

\[
\begin{align*}
2\alpha vv^T c_1 = & \frac{n}{n+2} E[q] c_1^T - \left( \delta_1 \hat{c}_1 v c_1^T + \nabla_x \cdot vv^T c_1^T - a \left( \tilde{\nabla}_x Q c_1^T + \nabla_x Q \nabla_x Q^T \right) \right) \\
\Rightarrow (2a+1)vv^T c_1 = & \frac{n}{n+2} E[q] c_1^T - \left( \delta_1 \hat{c}_1 v c_1^T + \nabla_x \cdot vv^T c_1^T - a \left( \tilde{\nabla}_x Q c_1^T + \nabla_x Q \nabla_x Q^T \right) \right) + \frac{a}{a+1} \nabla_x \cdot \nabla_x Q \left( \nabla_x Q \nabla_x Q^T \right)
\end{align*}
\]

where \( C_1 \) is a constant depending on \( a \) and \( n \).

### 3.3.1 Parabolic scaling

Let \( \kappa = 2 \) and \( E[q] = 0 \), \( T[q] = 0 \). Then (3.44) simplifies to

\[
\begin{align*}
\frac{\partial vv^T c_1}{\partial t} = & \frac{n+1}{n+1} E[q] c_1^T \\
\end{align*}
\]

Plugging (3.45) into (3.21) we arrive at an equation for \( c_1^T \):

\[
\begin{align*}
(a+1)\hat{c}_1 c_1 = & \frac{n+1}{n+1} \nabla_x \nabla_x^T : \left( E[q] c_1^T \right) - a \nabla_x \cdot \left( \nabla_x Q \nabla_x Q^T \right) + a \nabla_x \cdot \left( \nabla_x Q \nabla_x Q^T \right) \\
\end{align*}
\]

Combining (3.18) and (3.46) and using (3.4), we obtain for \( c_{01}^T \) the equation

\[
\begin{align*}
(a+1)\hat{c}_1 c_{01} = & \frac{n}{n+1} \nabla_x \nabla_x^T : \left( E[q] c_{01}^T \right) - a \nabla_x \cdot \left( \nabla_x Q \nabla_x Q^T \right) + a \nabla_x \cdot \left( \nabla_x Q \nabla_x Q^T \right) + O \left( \varepsilon^2 \right) \text{ if } \kappa = 2 \quad \text{and} \quad \frac{E[q]}{T[q]} = 0, \quad T[q] = 0.
\end{align*}
\]

**Remark 3.8.** A first order correction is often neglected in parabolic scaling. Here it allows to get a description of the taxis with respect to the macroscopic quantity \( Q \) that is more accurate than in (3.18). It turns out to be closer to a flux-limited taxis. The myopic diffusion remains unchanged.

**Remark 3.9.** Both conditions \( E[q] = 0 \) and \( T[q] = 0 \) are automatically satisfied if the tissue fibers are undirected (see Remark 3.5).

### 3.3.2 Hyperbolic scaling

Let \( \kappa = 1 \). Plugging (3.17) into (3.22) we arrive at the equation

\[
\begin{align*}
(a+1)\hat{c}_1 c_1 + \frac{n}{n+1} \nabla_x \cdot \left( E[q] c_1^T \right) = & -\hat{c}_1 c_1^T + \frac{n}{n+1} \nabla_x \nabla_x^T : \left( E[q] c_1^T \right) - a \nabla_x \cdot \left( \nabla_x Q \nabla_x Q^T \right) \text{ if } \kappa = 1
\end{align*}
\]

Utilising (3.19) twice we compute

\[
\begin{align*}
-\hat{c}_1 c_1^T = & \frac{n}{a+1} \nabla_x \cdot \left( \hat{c}_1 c_1 T[q] \right) \\
= & - \frac{n^2}{(a+1)^2} \nabla_x \cdot \left( E[q] \nabla_x \cdot \left( \nabla_x T[q] \right) \right) \\
\end{align*}
\]

Plugging (4.49) into (4.48) we obtain that

\[
\begin{align*}
(a+1)\hat{c}_1 c_1 + \frac{n}{n+1} \nabla_x \cdot \left( E[q] c_1^T \right)
\end{align*}
\]
Consequently, in virtue of (3.19) we obtain that \( \bar{c}_{01} \) satisfies the equation
\[
(1 + 1)\bar{c}_{01} + \frac{n}{n + 1} \nabla \cdot (E[q] \bar{c}_{01}) = \varepsilon \left( \frac{1}{2a + 1} \frac{n}{n + 2} \nabla \cdot (D[q] \nabla \bar{c}_{01}) - \frac{1}{(a + 1)^2} \frac{n}{n + 2} \nabla \cdot (E[q] \nabla \cdot (\bar{c}_{01} E[q])) \right) - a \nabla \cdot \left( \bar{c}_{01} \nabla_x Q \right). \tag{3.50}
\]

Remark 3.10. Typically for the hyperbolic case, the first order correction includes two terms which depend on the mesoscopic fiber orientation distribution: a myopic diffusion (the same as in (3.47)) and yet another term contributing to cell diffusion as well as to transport. The resulting diffusion tensor is (up to multiplication by a constant) a linear combination of two nonnegative definite matrices:
\[
\nabla [q] + \left( 1 - \frac{(2a + 1)(n + 2)}{(a + 1)^2(n + 1)^2} \right) E[q] \otimes E[q].
\]
Since the coefficient before the second matrix is obviously a positive number for any \( a > 0 \) and \( n \in \mathbb{N} \), the diffusion tensor is nonnegative definite.

The final term in the second line of (3.51) describes taxis with respect to the macroscopic quantity \( Q \). As is the case for the parabolic limit equation (3.18), the resulting taxis is not flux-limited. Even higher order approximations are necessary in order to capture this effect properly.

Overall, the first order correction effects a (small) deviation from the drift towards the average orientation of tissue fibers.

Remark 3.11 (Approximation order).

1. Equation (3.51) shows that the error of the first order approximation is \( O(\varepsilon^2) \). This confirms the surmise stated in [25] in connection with the hyperbolic scaling performed for a closely related problem. There, however, a different approach which is based on the Chapman-Enskog expansion was used in order to approximate the mesoscopic cell density.

2. Differentiating (3.3) and (3.8) with respect to \( \varepsilon \) at \( \varepsilon = 0 \) several times and performing the necessary calculations in the same manner as above, one obtains equations for corrections of higher order.

3.4 A mesoscopic first order approximation

One known issue with the above approach relying on the first order approximation is that \( \bar{c}_{01} \) is not necessarily nonnegative even if \( \bar{c} \) is nowhere negative. To fix this drawback one could consider instead an approximation \( \tilde{c}_{01} \) which vanishes for \( v \in \bar{c}B_1 \) and solves the following equation:
\[
\varepsilon \tilde{c} \nabla \tilde{c}_{01} + \varepsilon \nabla \cdot \left( v \tilde{c}_{01} \right) - a \nabla \cdot \left( (v - v^*) \tilde{c}_{01} \right) + \tilde{c}_{01} = nq \tilde{c}_{01} = nq(\varepsilon^0 + \varepsilon \tilde{c}) \tag{3.52a}
\]
\[
\tilde{c}_{01} \big|_{t=0} = c_0. \tag{3.52b}
\]
Equation (3.52a) is obtained from the original KTE (3.3) upon replacing the integral term by its first order approximation. Our next Proposition shows that the two main properties any reasonable approximation of a mesoscopic density should have are satisfied: both positivity and the total mass are preserved under (3.52a).

Proposition 3.12. Let \( \tilde{c}_{01} \) be a solution to (3.52). Suppose that \( \bar{c}_{01} \) and \( c_0 \) are nonnegative. Then \( \tilde{c}_{01} \) is also nonnegative and satisfies
\[
\int_{R^n} \tilde{c}_{01} \, dx = \int_{R^n} c_0 \, dx, \tag{3.53}
\]
i.e. the total mass is preserved.
Proof. To begin with, we apply the method of characteristics to the transport equation (3.52a) which leads to the ODE system

\[
\begin{align*}
\varepsilon^{\kappa-1} \partial_\alpha x &= v, \\
\varepsilon^\kappa \partial_\alpha v &= -a(v - v^*_a(x)), \\
\varepsilon^\kappa \partial_\alpha c_{01} &= -(na - 1)c_{01}^{\infty} = q^\kappa c_{01}^{\infty}.
\end{align*}
\]  

(3.54a) (3.54b) (3.54c)

It is obvious from (3.54c) that if \( \overline{c_{01}} \) and \( c_0 \) are nonnegative, then \( \overline{c_{01}} \) is nonnegative as well. It remains to prove (3.53). Recall that the full mesoscopic density \( c^* \) solves the measure conserving KTE (3.3) (this follows by integration of (3.5) by parts over \( \mathbb{R}^n \) with respect to \( x \)), so that

\[
\int_{\mathbb{R}^n} \overline{c} dx = \int_{\mathbb{R}^n} \overline{c_0} dx.
\]

This and the definition of \( c_{01}^{\infty} \) entail

\[
\int_{\mathbb{R}^n} \overline{c_{01}} dx = \int_{\mathbb{R}^n} \overline{c_0} dx.
\]

(3.55)

Further, we integrate (3.52a) with respect to \( x \) and \( v \) over the whole space \( \mathbb{R}^n \times \mathbb{R}_1(0) \). Using partial integration and (3.55) we obtain an ODE for the total mass:

\[
\frac{d}{dt} \int_{\mathbb{R}^n} \overline{c_{01}} \, dx + \int_{\mathbb{R}^n} \overline{c_{01}} \, dx = \int_{\mathbb{R}^n} \overline{c_{01}} \, dx = \frac{\int_{\mathbb{R}^n} \overline{c_0} \, dx}{d}. \quad (3.56)
\]

Finally, we integrate the initial condition (3.52b) over the whole space and obtain

\[
\int_{\mathbb{R}^n} \overline{c_{01}} \, dx \big|_{t=0} = \int_{\mathbb{R}^n} \overline{c_0} \, dx. \quad (3.57)
\]

Solving (3.56)-(3.57) yields (3.53). \( \square \)

Remark 3.13. In the previous Subsection we have obtained DTEs for \( \overline{c_{01}} \) for both parabolic and hyperbolic scaling cases. Each of those macroscopic equations for \( \overline{c_{01}} \) can be solved numerically, so that (3.52a) can be regarded as a linear transport equation which needs to be solved in order to determine \( \overline{c_{01}} \). This seems to be a useful alternative to dealing directly with the KTE (3.52a), since the numerical handling of such equations is known to be more expensive.

4 Rigorous limit procedures

4.1 Functional spaces

We begin with some basic notation. Let \( O \) be a domain or a smooth manifold. We denote by \( C^k_b(O) \) the space of real-valued functions which are continuous and bounded together with their derivatives up to order \( k \). For \( k = 0 \) we write \( C_b(O) \). If \( O \) is compact, we suppress the index \( b \). For a set \( A \subset O \) we denote by \( C^k_b(A) \) the subset of \( C^k_b(O) \) which consists of functions which are compactly supported in \( A \). Similarly, \( C_b(O; X) \) for \( X \) a Banach space denotes the space of continuous and bounded maps between \( O \) and \( X \).

Now we introduce various spaces of measures. We denote by \( M(O) \) the Banach space of finite signed Radon measures in \( O \), while \( M_+(O) \) stands for the closed subset of positive Radon measures. As usual, we use the total variation as norm on \( M(O) \).

The product \( \varphi \mu \) of a Radon measure \( \mu \in M(O) \) and a function \( \varphi \in C_b(O) \) is understood in the usual sense: it is a Radon measure which satisfies

\[
\int_O \psi(v) \, d(\varphi \mu)(v) = \int_O \psi \varphi(v) \, d\mu(v) \quad \text{for all } \varphi \in C_c(O).
\]

We recall that due to the Riesz-Markov-Kakutani representation theorem \( M(O) \) is isometrically isomorphic to the continuous dual of the separable normed space \( C_c(O) \). Further, we extend the -operator which denotes the integration over \( \mathbb{R}_1(0) \) to the case of measures: for \( \mu \in M(\mathbb{R}_1(0)) \) let

\[
\overline{\mu} := \int_{\mathbb{R}_1(0)} \, d\mu(v). \quad (4.1)
\]
In order to simplify the notation, we identify a measure $q \in \mathcal{M}(S_1(0))$ with the measure $q \times (r^{n-1} \, dr) \in \mathcal{M}(\mathcal{B}_1(0))$.

We also make use of the Banach space

$$L_{w-*}^\infty(\mathbb{R}^+; \mathcal{M}(\mathcal{O})) := \{ \mu : \mathbb{R}^+ \to \mathcal{M}(\mathcal{O}) \text{ is weak-\* measurable and} \}
\begin{align*}
\|\mu\|_{L_{w-*}^\infty(\mathbb{R}^+; \mathcal{M}(\mathcal{O}))} := & \|\mu(\cdot)\|_{L^\infty(\mathbb{R}^+)} < \infty \}
\end{align*}$$

and its closed subset

$$L_{w-*}^\infty(\mathbb{R}^+; \mathcal{M}_+(\mathcal{O})) := \{ \mu : \mathbb{R}^+ \to \mathcal{M}_+(\mathcal{O}) \mid \mu \in L_{w-*}^\infty(\mathbb{R}^+; \mathcal{M}(\mathcal{O})) \}.$$ 

Thereby we identify functions which coincide a.e. in $\mathbb{R}^+$. It is known [12, sections 8.18.1-8.18.2] that $L_{w-*}^\infty(\mathbb{R}^+; \mathcal{M}(\mathcal{O}))$ is isometrically isomorphic to the continuous dual of the Bochner space $L^1(\mathbb{R}^+)$ via the duality pairing

$$\langle \mu, \varphi \rangle = \int_{\mathbb{R}^+} \varphi(t) \, (d\mu(t))(x) \, dt.$$ 

Since $L^1(\mathbb{R}^+)$ is separable, the Banach-Alaoglu theorem implies that balls in $L_{w-*}^\infty(\mathbb{R}^+; \mathcal{M}(\mathcal{O}))$ are weak-\* sequentially compact.

Finally, we introduce the spaces

$$C_{w-*}(\mathbb{R}^+; \mathcal{M}(\mathcal{O})) := \{ \mu : \mathbb{R}^+ \to \mathcal{M}(\mathcal{O}) \text{ is weak-\* continuous and} \}
\begin{align*}
\|\mu(\cdot)\|_{\mathcal{M}(\mathcal{O})} := & \|\mu(\cdot)\|_{L^\infty(\mathbb{R}^+)} < \infty \}
\end{align*}$$

and its closed subset

$$C_{w-*}(\mathbb{R}^+; \mathcal{M}_+(\mathcal{O})) := \{ \mu : \mathbb{R}^+ \to \mathcal{M}_+(\mathcal{O}) \mid \mu \in C_{w-*}(\mathbb{R}^+; \mathcal{M}(\mathcal{O})) \}.$$ 

### 4.2 Main results and their proofs

From now on we assume the model parameters to satisfy the following conditions:

**Assumptions 4.1.**

1. $q \in C_b(\mathbb{R}^n; \mathcal{M}_+(S_1(0)))$ and

$$\overline{q} = \frac{1}{n};$$

2. $Q \in C_b(\mathbb{R}_0^+ \times \mathbb{R}^n)$, $\nabla_x Q \in C_b(\mathbb{R}_0^+ \times \mathbb{R}^n; \mathbb{R}^n)$, $F \in C_b(\mathbb{R}^n; \mathbb{R}^{n \times n})$,

$$\|\|F\|\|_{C_b(\mathbb{R}^{n \times n})} < 1.$$

**Remark 4.2** (Moments of $q$). Due the regularity assumption on $q$ the moments

$$\mathbb{E}[q](x) := \int_{S_1(0)} \nu(v) (dq(x))(v),$$

$$\mathbb{D}[q](x) := \int_{S_1(0)} \nu(v) \nabla(v) (dq(x))(v)$$

are well-defined and satisfy

$$\mathbb{E}[q] \in C_b(\mathbb{R}^n; \mathbb{R}^n), \quad \mathbb{D}[q] \in C_b(\mathbb{R}^n; \mathbb{R}^n \times \mathbb{R}^n).$$

Next, we define weak measure-valued solutions to the KTE (3.3) and the limit equations.

**Definition 4.3** (Weak solutions to the KTE). Let $\varepsilon > 0$ and let Assumptions 4.1 be satisfied and let $\varepsilon_0 \in \mathcal{M}_+(\mathbb{R}^n \times \mathcal{B}_1(0))$ be some initial data. We call an element $\varepsilon \in L_{w-*}^\infty(\mathbb{R}^+; \mathcal{M}_+(\mathbb{R}^n \times \mathcal{B}_1(0)))$ a weak solution to (3.3) if for all $\varphi \in C^1_c(\mathbb{R}^n)$, $\psi \in C^1(\mathcal{B}_1(0))$, and $\eta \in C^1_c(0, \infty)$ it holds that

$$-\varepsilon \eta(0) \int_{\mathbb{R}^n \times \mathcal{B}_1(0)} \varphi \psi \, dc_0 - \int_0^{\infty} \int_{\mathbb{R}^n \times \mathcal{B}_1(0)} (\varepsilon \nabla(v - v_0) \cdot \nabla(v \varphi \psi)) \, dc_0 (t) \, dt$$
\[
\int_0^\infty \left( \sum_{n \in \mathbb{N}} \psi_n \varphi \, d\overline{c} - \int_{\mathbb{R}^n \times \mathbb{N}(0)} \varphi \psi \, dc(t) \right) \, dt,
\]
and the total mass is preserved:
\[
\int_{\mathbb{R}^n} dc = \int_{\mathbb{R}^n} \overline{c}_0 \quad \text{a.e. in } (0, \infty).
\]

**Remark 4.4** (Solvability). In this work we are mostly interested in upscaling and do not deal with solvability of KTEs. These equations have been studied by many authors, though mostly in the physical context. A sketch of the proof of the existence of solutions to (2.13) is provided in Appendix A for the reader’s convenience.

**Definition 4.5** (Weak solutions to the parabolic limit). Let Assumptions 4.1 be satisfied and let \( \overline{c}_0 \in \mathcal{M}_+(\mathbb{R}^n) \) be some initial data. We call an element \( \overline{c}(t) \in L^\infty_{w, a}(\mathbb{R}^n; \mathcal{M}_+(\mathbb{R}^n)) \) a weak solution to (3.18) if for all \( \varphi \in C^2_b(\mathbb{R}^n) \) and \( \eta \in C^1_b[0, \infty) \) it holds that
\[
-\eta(0) \int_{\mathbb{R}^n} \varphi \, d\overline{c}_0 - \int_0^\infty \left( \frac{dn}{dt} \int_{\mathbb{R}^n} \varphi \, d\overline{c}(t) + \eta \int_{\mathbb{R}^n} \nabla_x \varphi \cdot \left( \frac{1}{a + 1} \mathbb{E}[q] \, d\overline{c}(t) \right) \right) \, dt = 0.
\]

**Definition 4.6** (Weak solutions to the hyperbolic limit). Let Assumptions 4.1 be satisfied and let \( \overline{c}_0 \in \mathcal{M}_+(\mathbb{R}^n) \) be some initial data. We call an element \( \overline{c}(t) \in L^\infty_{w, a}(\mathbb{R}^n; \mathcal{M}_+(\mathbb{R}^n)) \) a weak solution to (3.19) if for all \( \varphi \in C^2_b(\mathbb{R}^n) \) and \( \eta \in C^1_b[0, \infty) \) it holds that
\[
-\eta(0) \int_{\mathbb{R}^n} \varphi \, d\overline{c}_0 - \int_0^\infty \left( \frac{dn}{dt} \int_{\mathbb{R}^n} \varphi \, d\overline{c}(t) + \eta \int_{\mathbb{R}^n} \nabla_x \varphi \cdot \left( \frac{1}{a + 1} \mathbb{E}[q] \, d\overline{c}(t) \right) \right) \, dt = 0.
\]

**Remark 4.7.** Due to the assumptions made on \( q, Q, \mathbb{F}, \) and \( \mathbb{C}_0 \), each of the integrals in (4.5)-(4.8) is well-defined and finite.

Thus defined weak solutions satisfy in a weak sense the equations for the moments which were formally derived in Section 3.

**Proposition 4.8** (Weak moment equations). Let \( \overline{c}(t) \) be a weak solution as in Definition 4.3. Then equations (3.5) and (3.8) are satisfied, respectively, in the following weak sense: for all \( \varphi \in C^1_b(\mathbb{R}^n) \) and \( \eta \in C^1_b[0, \infty) \)
\[
-\eta(0) \int_{\mathbb{R}^n} \varphi \, d\overline{c}_0 - \int_0^\infty \left( \frac{dn}{dt} \int_{\mathbb{R}^n} \varphi \, d\overline{c}(t) + \epsilon^{1-\kappa} \eta \int_{\mathbb{R}^n} \nabla_x \varphi \cdot dc(t) \right) \, dt = 0,
\]
and for all \( \varphi \in C^2_b(\mathbb{R}^n) \) and \( \eta \in C^2_b[0, \infty) \)
\[
-\epsilon^{1-\kappa} \frac{1}{a + 1} \frac{d\eta}{dt} \int_{\mathbb{R}^n} \varphi \, d\overline{c}_0 - \epsilon^{1-\kappa} \frac{1}{a + 1} \int_0^\infty d^2 \eta \int_{\mathbb{R}^n} \varphi \, d\overline{c}(t) \, dt - \eta(0) \int_{\mathbb{R}^n} \varphi \, d\overline{c}_0 - \epsilon^{1-\kappa} \frac{1}{a + 1} \int_0^\infty d\eta \int_{\mathbb{R}^n} \varphi \, d\overline{c}(t) \, dt
\]

\[
= \int_0^\infty \eta \int_{\mathbb{R}^n} \nabla_x \varphi \cdot \epsilon^{2-\kappa} \frac{1}{a + 1} \mathbb{E}[q] \, d\overline{c}(t) \, dt
\]

\[
+ \int_0^\infty \eta \int_{\mathbb{R}^n} \nabla_x \varphi \cdot \epsilon^{1-\kappa} \frac{a}{a + 1} d\overline{c}(t) \, dt
\]

\[
- \eta(0) \int_{\mathbb{R}^n} \nabla_x \varphi \cdot \epsilon^{1-\kappa} \frac{a}{a + 1} \, d\overline{c}_0.
\]

**Proof.** Taking \( \psi \equiv 1 \) in (4.5) directly implies (4.9). Similarly, using \( \psi(v) = v \) and \( \nabla_x \varphi \) in place of \( \varphi \) for some \( \varphi \in C^2_b(\mathbb{R}^n) \), we obtain from (4.5) that
\[
\int_0^\infty \epsilon^{1-\kappa} \eta \int_{\mathbb{R}^n} \nabla_x \varphi \cdot d\overline{c}(t) \, dt
\]

\[
= \int_0^\infty \eta \int_{\mathbb{R}^n} \nabla_x \varphi \cdot \epsilon^{2-\kappa} \frac{1}{a + 1} \mathbb{E}[q] \, d\overline{c}(t) \, dt
\]

\[
+ \int_0^\infty \eta \int_{\mathbb{R}^n} \nabla_x \varphi \cdot \left( \epsilon^{1-\kappa} \frac{a}{a + 1} \, d\overline{c}(t) + \epsilon^{1-\kappa} \frac{1}{a + 1} \mathbb{E}[q] \, d\overline{c}(t) \right) \, dt
\]
Further, we plug \( \frac{d\eta}{dt} \) instead of \( \eta \) for some \( \eta \in C^2[0, \infty) \) into (4.9) and obtain that
\[
-\frac{d\eta}{dt}(0) \int_{\mathbb{R}^n} \varphi \, d\vec{c}_0 - \int_0^\infty \int_{\mathbb{R}^n} \varphi \, d\vec{c}(t) + \varepsilon^{1-\kappa} \frac{d\eta}{dt} \int_{\mathbb{R}^n} \nabla_x \varphi \cdot d\vec{c}(t) dt = 0
\]  
(4.12)

Combining (4.9), (4.11), and (4.12) we arrive at (4.10).

Now we can state and prove the following upscaling result:

**Theorem 4.9** (Parabolic limit). Let Assumptions 4.1 be satisfied. Assume that \( \kappa = 2 \) and \( E[q] = 0 \). For some \( \varepsilon_m \to 0 \) let \( c^{\varepsilon_m} = M_+([\mathbb{R}^n \times \overline{B}_1(0)]) \) be a sequence of initial data such that
\[
cr_{0}^{\varepsilon_m} \to c_0 \text{ in } M_+([\mathbb{R}^n \times \overline{B}_1(0))
\]
(4.13)
for some \( c_0 \in M_+([\mathbb{R}^n \times \overline{B}_1(0)) \). Finally, let \( c^{\varepsilon} \) be a weak solution to (3.3) in terms of Definition 4.3 corresponding to \( c^{\varepsilon_m} \). Then there exists a subsequence \( c^{\varepsilon_{m_k}} \) such that
\[
cr^{\varepsilon_{m_k}} \to c_0 \text{ in } L_{w,*}^\infty([\mathbb{R}^+; M_+([\mathbb{R}^n \times \overline{B}_1(0))])
\]
(4.14)
where \( c_0 \) satisfies (3.10) and \( c_0 \) is a weak solution to (3.18) in terms of Definition 4.5 corresponding to \( c_0 \).

**Remark 4.10.** Other than in [24] where the parabolic limit involved just self-diffusion of the cell population with a similar diffusion coefficient \( D[q] \), we do not require \( q \) to be constant with respect to \( x \). Such relaxation of conditions imposed on \( q \) is highly relevant from the application viewpoint, since the orientational distribution of tissue fibers usually varies from one point in space to the other. The spacial heterogeneity of the tissue structure plays a major role in haptotactic behaviour.

**Proof.** (of Theorem 4.9) To begin with, we observe that since the sequence of initial measures is weak-* converging, it is also uniformly bounded. Consequently, the mass preservation property (4.6) implies that \( c^{\varepsilon_m} \) is uniformly bounded in \( L^\infty_{w,*}([\mathbb{R}^+; M_+([\mathbb{R}^n \times \overline{B}_1(0))]) \). Thus, there exists a subsequence \( c^{\varepsilon_{m_k}} \) weak-* converges to some \( c^0 \in L^\infty_{w,*}([\mathbb{R}^+; M_+([\mathbb{R}^n \times \overline{B}_1(0))]) \). This allows to pass to the limit for \( \varepsilon = \varepsilon_{m_k} \) as \( k \to \infty \) in equation (4.5) from Definition 4.3. Thus, we obtain that
\[
\int_0^\infty \eta \left( \int_{\mathbb{R}^n \times \overline{B}_1(0)} av \cdot \eta \varphi \psi \, dc^0(t) \right) dt
= \int_0^\infty \eta \left( n \int_{\mathbb{R}^n} \varphi \int_{\overline{B}_1(0)} \psi \, dq \, dc^0(t) - \int_{\mathbb{R}^n \times \overline{B}_1(0)} \varphi \psi \, dc^0(t) \right) dt.
\]
(4.15)

Resolving this distributional equation we conclude that
\[
\int_{\overline{B}_1(0)} av \cdot \nabla_x \psi \, dc^0 = n \int_{\overline{B}_1(0)} \psi \, dq - \int_{\overline{B}_1(0)} \psi \, dc^0 \text{ in } L^\infty_{w,*}([\mathbb{R}^+; M_+([\mathbb{R}^n])).
\]
(4.16)

In particular, taking \( \psi(v) = v \psi\) in (4.16) we obtain the expression (3.17) for the second moment of \( c^0 \). Due to Proposition 4.8, for each \( m \in \mathbb{N} \) the weak solution \( c^{\varepsilon_m} \) satisfies the moment equation (4.10). Passing to the limit in this equation, we obtain that
\[
-\eta(0) \int_{\mathbb{R}^n} \varphi \, dc^0 - \int_0^\infty \frac{d\eta}{dt} \int_{\mathbb{R}^n} \varphi \, dc(t)
= \int_0^\infty \eta \left( \int_{\mathbb{R}^n} \nabla_x \varphi \cdot \frac{1}{a+1} d\eta \right) + \int_{\mathbb{R}^n} \nabla_x \varphi \cdot \frac{a}{a+1} F \nabla_q Q \, dc(t) dt.
\]
(4.17)

Substituting (3.17) into (4.17), we arrive at the weak formulation (4.7) from Definition 4.5.

For a fixed \( c^0 \) a solution to (4.16) is given by (3.10). Since the equation is a linear one, this solution is also unique if the corresponding homogeneous equation
\[
\int_{\overline{B}_1(0)} (av \cdot \nabla_x \psi + \psi) \, du = 0 \text{ for all } \psi \in C^1(\overline{B}_1(0)).
\]
(4.18)
has only the trivial solution in \( B_1(0) \). Observe that each \( g \in C_c(\overline{B}_1(0) \setminus \{0\}) \) can be described as

\[
av \cdot \nabla v \psi + \psi = g,
\]

where

\[
\psi_y(v) = -|v|^\frac{\kappa}{p} \int_0^1 g(y \text{sign}(v)) \, dy
\]

belongs to \( C^1(\overline{B}_1(0)) \). Consequently, \( \text{supp } u \subset \{0\} \), i.e. \( u = C\delta_0 \) where \( \delta_0 \) denotes the Dirac delta and \( C \) is some constant. But then

\[
0 = \int_{\overline{B}_1(0)} (av \cdot \nabla v \psi + \psi) \, du = C\psi(0) \quad \text{for all } \psi \in C^1(\overline{B}_1(0)),
\]

which implies that \( C = 0 \). This shows that \( u = 0 \).

An analogous result holds for the hyperbolic case:

**Theorem 4.11** (Hyperbolic limit). Let Assumptions 4.1 be satisfied. Assume that \( \kappa = 1 \). For some \( \epsilon_m \to 0 \) let \( c^{\epsilon_m}_0 \in M_+(\mathbb{R}^n \times B_1(0)) \) be a sequence of initial data such that

\[
c^{\epsilon_m}_0 \xrightarrow{m \to \infty} c^0_0 \quad \text{in } M_+(\mathbb{R}^n \times \overline{B}_1(0))
\]

for some \( c^0_0 \in M_+(\mathbb{R}^n \times \overline{B}_1(0)) \). Finally, let \( c^{\epsilon_m} \) be a weak solution to (3.3) in terms of Definition 4.3 corresponding to \( c^{\epsilon_m}_0 \). Then there exists a subsequence \( c^{\epsilon_{m_k}} \) such that

\[
c^{\epsilon_{m_k}} \xrightarrow{m_k \to \infty} c^0 \quad \text{in } L^\infty_{\text{loc}}(\mathbb{R}^+; M_+(\mathbb{R}^n \times \overline{B}_1(0))),
\]

where \( c^0 \) satisfies (3.10) and \( c^0 \) is a weak solution to (3.19) in terms of Definition 4.6 corresponding to \( c^0_0 \).

We omit the proof of this Theorem since it is very similar to that of Theorem 4.9.

**Remark 4.12.** In the same way as we have proved the rigorous convergence for the zero order approximations one could validate the other formal derivations performed in Section 3.

## 5 Discussion and outlook

In recent years modelling with KTEs in the multiscale modelling framework has proved to be an effective approach to describing cell movement in a fibrous environment, as it carefully connects single cell dynamics with the evolution of one or several cell distribution functions depending on time, position, velocity, and possibly further activity variables. Since such mesoscopic models are generally difficult to handle numerically, suitable macroscopic approximations, such as, e.g., limits of parabolic or hyperbolic scalings, are often derived and solved instead. In this work we developed a new approach to dealing with such scalings for a general class of KTEs involving transport with respect to velocity. It relies on the method of characteristics and a differential equation (3.8) which connects moments of zero and second order. The latter key equation can be utilised for both parabolic and hyperbolic scalings and offers a unified and transparent way of deriving macroscopic equations for approximations of an arbitrary high order. As an illustration, we have deduced DTEs for zero and first order approximations on the macroscale for both mentioned scaling types for our KTE. It turns out that our formal computations can be mimicked by the corresponding operations with Radon measures. Under rather general conditions on the parameters which allow for a spatially heterogeneous measure-valued fiber orientation distribution, we have thus been able to validate our limit passages rigorously.

Most of the previous constructions leading from KTEs to RDTEs rely on a Hilbert space structure already on the level of formal scalings. Indeed, one typically assumes the zero and first order approximations to be orthogonal in a particular weighted space of square integrable functions. We have actually seen that this property fails to hold for the model class considered here. Conversely, our approach does not rely on orthogonality in any way and is applicable to a broader class of KTEs.

In addition to the macroscopic approximations of zero order or higher we have also developed a transport equation which preserves positivity and the total mass and can be solved numerically in order to obtain a suitable mesoscopic first order approximation of the solutions to the original KTE.
The class of KTEs we have used here to illustrate our approach can account for a number of motility features. On the macroscale, it has led to DTEs which, depending on the chosen scaling type and approximation order, include such terms as: myopic diffusion, drift, and taxis with respect to a mesoscopic and/or macroscopic quantity. The latter macroscopic quantity could be a haptot- or a chemoattractant. In our model, the taxis with respect to such an attractant is caused by biochemical and/or biophysical stress perceived by the cells. It is modelled via Newton’s second law in (2.3b) and includes flux-limitation. We have seen that flux-limited taxis can be recovered on the macroscale provided that an approximation of a sufficiently high order is used. One way to extend our model would be to consider a more general form of acceleration in (2.3b), e.g. by letting the acceleration scaling coefficient depend on the attractant. A similar consideration was made in [6] regarding the so-called ‘chemotaxis force’. One could also consider dependencies on other macroscopic quantities, including cell population density. The latter, however, would require dealing with convergences in nonlinear terms and would render a rigorous limit passage considerably more difficult. This is because the weak type of convergence used in this work would no longer be sufficient in order to handle nonlinearities. Further, we have assumed the attractant to be some given function, thus allowing us to deal with a single equation accounting for the cell motion. A more realistic model would have to include an equation characterising the dynamics of the attractant. That would be an ODE if the tactic cue represents, for instance, volume fraction of tissue which is supposed to be degraded by tumour cells. If, on the other hand, it represents the concentration of some chemoattractant, then we would need to consider a reaction-diffusion PDE with source terms characterising production by tumour cells depending on their local macroscopic density and decay caused by other influences, along with a linear diffusion. In both cases, however, we would then be dealing with a nonlinear, strongly coupled system. Once again this would make the rigorous analysis much more challenging, if at all possible.

All motility terms obtained in the macroscopic formulations carry some information about the underlying tissue structure: the drift and diffusion coefficients depend on the orientation distribution of tissue fibers, whereas the chemotactic sensitivity tensor which controls the taxis with respect to a macroscopic attractant accounts for spatial heterogeneity. For simplicity we have taken the fiber distribution to be time-independent, assuming it to be some given function of spatial position and orientation. Relaxing this assumption would lead (for an example of a formal, orthogonality-based deduction and numerical simulations of the obtained equations see [9]) to a nonlinear strongly coupled meso-macro system with dynamically changing diffusion coefficient and drift velocity which is highly challenging from the analytical point of view.

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Assumptions

Let us first assume that in addition to conditions 1.-2. from Assumptions 4.1, we have

\[ \nabla_x \nabla_T Q \in C_b(\mathbb{R}_0^+ \times \mathbb{R}^n; \mathbb{R}^{n \times n}), \quad \nabla F \in C_b(\mathbb{R}^n; \mathbb{R}^{n \times n \times n}). \]

Under the assumptions we made, standard ODE theory implies that the ODE system (2.3) is globally uniquely solvable on \( \mathbb{R}^n \times \mathbb{R}^n \). We denote by \( Z(t, s) : \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}^n \times \mathbb{R}^n \) the corresponding solution operator, meaning that \( Z(t, s)(z_0) \) is the value at time \( t \) of the solution of (2.3) which has started in \( z_0 \) at the initial time \( s \). Again, the ODE theory implies that this map is well-defined and bijective for all \( s, t \in \mathbb{R} \). Moreover, exploiting the form of \( S \), we see that

\[ (Z(t, s))^{-1}(\mathbb{R}^n \times (\mathbb{R}^n \setminus B_1(0))) \subset \mathbb{R}^n \times (\mathbb{R}^n \setminus \overline{B}_1(0)) \quad \text{for } t > s. \]

Let us set

\[ c_0 := 0, \quad q := 0 \quad \text{in } \mathbb{R}^n \times (\mathbb{R}^n \setminus \overline{B}_1(0)), \]

so that

\[ c_0 \in \mathcal{M}_+(\mathbb{R}^n \times \mathbb{R}^n), \quad q \in C_b(\mathbb{R}^n; \mathcal{M}_+(\mathbb{R}^n)). \]
We recall that the Cauchy problem for the conservative transport equation
\[ \partial_t \mu + \nabla_z \cdot (V(t) \mu) = 0 \]  \hspace{1cm} (A.3)
has a unique weak solution \( \mu \in C_{w,*}(\mathbb{R}^+; \mathcal{M}_+([\mathbb{R}^n \times \mathbb{R}^n])) \) for every initial value \( \mu_0 \in \mathcal{M}_+([\mathbb{R}^n \times \mathbb{R}^n]) \), see e.g. [38, Chapter 5 Theorem 5.34]. Moreover, its solution is given by an explicit formula:
\[ \mu(t) = Z(t,0)\#\mu_0, \]  \hspace{1cm} (A.4)
where \( T\#\mu \) denotes the push-forward of measure \( \mu \) under a map \( T : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}^n \times \mathbb{R}^n \). We recall that the Cauchy problem for the conservative transport equation
\[ \partial_t \mathbb{P} + \nabla_z \cdot (V(t) \mathbb{P}) = 0 \]  \hspace{1cm} (A.5)
has a unique weak solution \( \mathbb{P} \in C_{w,*}(\mathbb{R}^+; \mathcal{M}_+([\mathbb{R}^n \times \mathbb{R}^n])) \) for every initial value \( \mathbb{P}_0 \in \mathcal{M}_+([\mathbb{R}^n \times \mathbb{R}^n]) \), see e.g. [38, Chapter 5 Theorem 5.34]. Moreover, its solution is given by an explicit formula:
\[ \mathbb{P}(t) = Z(t,0)\#\mathbb{P}_0, \]  \hspace{1cm} (A.6)
where \( T\#\mathbb{P} \) denotes the push-forward of measure \( \mathbb{P} \) under a map \( T : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}^n \times \mathbb{R}^n \). We recall that the Cauchy problem for the conservative transport equation
\[ \partial_t \mathbb{Q} + \nabla_z \cdot (V(t) \mathbb{Q}) = 0 \]  \hspace{1cm} (A.7)
Combining (A.3)–(A.4), we compute: for all \( t \in \mathbb{R}^+ \)
\[ \mathbb{P}(t) = \mathbb{P}(0) + \int_0^t \mathbb{P}(s) - \mathbb{P}(t) \, ds \]  \hspace{1cm} (A.8)
\[ \text{as required.} \]

Next, we verify that solutions to (A.6) are weak solutions to the KTE (2.13). Using (A.3), (A.4), and (A.6), we compute: for all \( t \in \mathbb{R}^+ \)
\[ \mathbb{Q}(t) = \mathbb{Q}(0) + \int_0^t \mathbb{Q}(s) - \mathbb{Q}(t) \, ds \]  \hspace{1cm} (A.9)

Combining (A.7)–(A.9), we conclude that the solution to (A.6) is a weak solution to the KTE (2.13) in terms of Definition 4.3. A standard approximation argument finally yields existence under the original Assumptions 4.1.

**Remark A.2** (Uniqueness). Under some additional smoothness of \( Q \) and \( q \) with respect to \( x \) variable one can verify the uniqueness of weak solutions to (2.13).