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Citation for the original published paper (version of record):

Eden, M., Nikolopoulos, C., Muntean, A. (2021)

"A multiscale quasilinear system for colloids deposition in porous media: Weak solvability and numerical simulation of a near-clogging scenario *Nonlinear Analysis: Real World Applications* 

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# A multiscale quasilinear system for colloids deposition in porous media: Weak solvability and numerical simulation of a near-clogging scenario

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August 8, 2021

### **Abstract**

We study the weak solvability of a macroscopic, quasilinear reaction-diffusion system posed in a 2D porous medium which undergoes microstrucural problems. The solid matrix of this porous medium is assumed to be made out of circles of not-necessarily uniform radius. The growth or shrinkage of these circles, which are governed by an ODE, have direct feedback to the macroscopic diffusivity via an additional elliptic cell problem.

The reaction-diffusion system describes the macroscopic diffusion, aggregation, and deposition of populations of colloidal particles of various sizes inside a porous media made of prescribed arrangement of balls. The mathematical analysis of this two-scale problem relies on a suitable application of Schauder's fixed point theorem which also provides a convergent algorithm for an iteration method to compute finite difference approximations of smooth solutions to our multiscale model. Numerical simulations illustrate the behavior of the local concentration of the colloidal populations close to clogging situations.

**Keywords:** Colloidal transport and deposition, reactive porous media, weak solutions to strongly nonlinear parabolic systems, two-scale finite difference approximation, clogging

MSC2020: 35K61, 65N06, 35B27, 76S05, 80M40

# 1 Introduction and problem statement

We study a two-scale system modeling the effective diffusive transport as well as the aggregation and deposition of populations of colloidal particles inside porous media. Such situations arise, for instance, in membrane filtration scenarios [13, 34], papermaking [29], immobilization of colloids in soils [7], or transport of colloidal contaminants in groundwater [41].

We are particularly interested in situations where micro-structural changes due to the adsorption or desorption of colloids are allowed to take place. This can locally change both the transport patterns and storage capacity of the medium; see [4, 10, 15, 23, 36, 44] for related cases. This variety of technological and natural processes is based on the transfer of colloidal particles from liquid suspension onto stationary surfaces [20]. From this perspective, one can perceive that the porous media we are considering here behave like materials with reactive internal microstructures (see [8] for a periodic setting) and, based on [39], they are sometimes classified as media with distributed microstructures. Additional motivation for this work comes from our own research on reactive flow in porous media and is linked very much with the work of P. Ortoleva and J. Chadam (see e.g. [6] and follow up papers), but it is worth mentioning that quite related aspects arise in pharmacy and

medicine like drug delivery, thrombosis formation on arterial walls, evolution of Alzheimer's disease. We refer the reader, for instance, to [27, 5, 43, 14] for works in this direction.

Denoting with  $u=(u_1,...,u_N)$  (i=1,...,N) the molar concentrations of colloids of size i (with  $N\in\mathbb{N}$  the maximal size), its time evolution can be modelled by a quasi-linear parabolic system in the form of

$$\partial_t u_i - \operatorname{div}(\widehat{D}_i(u)\nabla u_i) = F_i(u),$$
 (1)

where  $F_i(u)$  accounts for the aggregation, segregation, and adsorption processes and  $\widehat{D}_{ii}(u_i)$  the changing diffusivity as a consequence of the micro-structural changes (like clogging) inside the porous medium itself. While equation (1) is purely macroscopic, the computation of the effective diffusivity  $D_i(u_i)$  is done on the micro-scale therefore leading to the two-scale nature of our problem. This system is a compact and abstract reformulation of a two-scale model for colloidal transport derived in [26] via asymptotic homogenization (more details are given in Section 2). Structurally similar (two-scale model with geometrical changes) models were investigated in, e.g., [12, 32].

In this work, we take a 2D-cross-section of a porous medium and assume the solid matrix of the cross section to be made up of circles of not-necessarily uniform radius. The growth and shrinkage of these circles, which represent the underlying micro-structural changes of the porous medium, are modelled via a scalar quantity governed by an additional ODE. For a similar geometrical setup see, e.g., [32]. The model and the resulting mathematical problem gets more complicated if we were to allow for more general geometries (e.g., evolving  $C^2$ -interfaces) that can not be represented by a scalar quantity like the radius in our setting. We treat our geometries in 2D mainly for the sake of simplicity of inequalities and transformations and also because the simulation work is easier to be handled in 2D compared to 3D, there is no fundamental element in the analysis that is sensitive to dimensions (like e.g. Sobolev embeddings would be). As a consequence, the mathematical analysis part can be extended to 3D with suitable modifications on the upper and lower *a priori* bounds on the radii of the balls-like microstructure.

The quasilinear structure of the problem together with the multiscale coupling is non-standard. Here, we point out that  $\widehat{D}_i$  and  $F_i$  are non-linear operators that are not defined via point wise evaluation (in the sense of  $\widehat{D}_i(u)(t,x)=\widehat{D}_i(t,x,u(t,x))$ ). In particular, it does not fit directly to the framework elaborated in, e.g., [2] and it requires an approach that utilizes the underlying coupling present in the model equations behind the abstract system. A similar two-scale problem allowing for micro-structural changes was investigated in [24].

In Section 2 we explain our working strategy to prove the existence of weak solutions to the overall problem. To keep things simple, we consider only the case where the diffusivities  $\widehat{D_i}$  do not degenerate. Note however, that it is technically possible to include simple degeneracies in the analysis of these types of problems (like neighboring microstructures touching in single points [38]), a complete (local) clogging being however out of reach. Besides the non-degeneracy of the effective parameter, another simplification is included – the absence of the flow. Note that if the colloidal populations would involve Janus particles immersed in a fluid flow, then, most likely, besides the balance equations of the linear momentum one would also have to take into account also the Nernst-Planck production and subsequent transport of electrical charges; see e.g. [19, 35] for more information in this direction.

The paper is organized as follows: In Section 2 we present the model and outline our strategy for the analysis of our problem. We list the needed mathematical details of the problem so that we can prove in Section 3 the existence of a weak solution. In Section 4, we solve numerically our multiscale quasilinear problem and discuss the obtained numerical results for realistic parameter regimes. We add in Section 5 a detailed discussion of the potential of our problem, expected results, and related aspects.

# 2 Problem statement and solution strategy

In the following, let S=(0,T) be the time interval of interest and  $\Omega\subset\mathbb{R}^2$  a bounded Lipschitz domain with outer unit normal vector n=n(x).

In addition, let  $N \in \mathbb{N}$  be a given number indicating the maximal possible size of an aggregate of colloid particle, where size refers to the number of primary particles making up the aggregate. For each  $i = \{1, ..., N\}$ , let  $u_i \colon S \times \Omega \to [0, \infty)$  (we set  $u = (u_1, ..., u_N)$ ) denote the molar concentration of aggregates of size i at point  $x \in \Omega$  at time  $t \in S$ . We take the function  $v \colon S \times \Omega \to [0, \infty)$  to represent the mass density of absorbed material (mass that is in the system but currently not part of the diffusion and agglomeration process); this mass can be dissolved again allowing colloidal populations to re-enter the pore space. This process of absorption and desorption is modelled in this context via an Robin-type exchange term in the form of Henry's law (see e.g. [16, 22]) with positive exchange coefficients  $a_i$  and  $b_i$ :

$$\frac{2\pi r}{1-\pi r^2}(a_iu_i-b_iv).$$

Here, the radius function  $r\colon S\times\Omega\to(0,r_{max})$  (for some  $r_{max}>0$ ) acts as a measure of the clogginess of the porous media (see Figure 1). Note that  $r_{max}$  will be chosen (the details are outlined in Section 3) such that actual clogging scenarios are excluded; that is neighboring balls are not allowed to touch. The exchange coefficient  $\frac{2\pi r}{1-\pi r^2}$  is given as the ratio of the perimeter of the ball  $(2\pi r)$  and the pore volume; this relationship is justified via the upscaling in [26]. Note that equilibrium between  $u_i$  and v is accomplished when  $u_i=a_i/b_iv$ .

To describe the aggregation processes taking place inside the pore space of the medium, we use a truncated variant of the *Smoluchowski* formulation (we point to [1] for a review) given here by

$$R_{i}(u) = \frac{1}{2} \sum_{j+l=i} \gamma_{jl} u_{j} u_{l} - u_{i} \sum_{j=1}^{N-i} \gamma_{ij} u_{j}$$
 (2)

with positive coefficients  $\gamma_{il}$ .

**Remark 1.** The ansatz given by equation (2) is a truncated version of the discrete Smoluchowski coagulation term (see, e.g., [1])

$$R_i(u) = \frac{1}{2} \sum_{j+l=i} \gamma_{jl} u_j u_l - u_i \sum_{j=1}^{\infty} \gamma_{ij} u_j$$

where we only consider colloids up to a maximal size of N>0 (i.e.,  $\gamma_{jl}=0$  for j+l>N). The first sum in equation (2) accounts for the formation via coagulation under the assumption that colloids of size i can be formed when two smaller colloids of sizes j and l with the property j+l=i meet. In particular, the first sum is zero in  $R_1$  (particles of size 1 can not form via coagulation) and  $1/2\gamma_{11}u_1^2$  in  $R_2$  (particles of size 2 can only form via coagulation of two size-1 colloids). The second sum accounts for the loss of i-sized colloids by coagulating with a different colloid of size j to form a new one of size  $i+j\leq N$ . The second sum is therefore empty in  $R_N$ . Note that this is conceptually very similar to viewing the overall production process as a chain of second order chemical reactions. The truncation to a fixed population size holds particularly in the case of colloids in soils. If no truncation is taken into account, then fragmentation terms are needed to prevent the size of the population to grow indefinitely. In that case, the functional setting changes as for instance in [18] and references cited therein.

It is important to note that in the context of porous media the colloidal populations involve a finite size chain of the cluster, i.e. there will be a population of N-mers where N takes the maximum cluster size. For that reason, we deal with a truncated finite sum here. Interestingly, for many applications a good choice of such N is rather low; see e.g. [22].

The diffusion-reaction system for the different aggregates is then given via

$$(P_i) \begin{cases} \partial_t u_i - \operatorname{div}(D_i(r)\nabla u_i) = R_i(u) - \frac{2\pi r}{1 - \pi r^2}(a_i u_i - b_i v) & \text{in } S \times \Omega, \\ -D_i(r)\nabla u_i \cdot n = 0 & \text{on } S \times \partial \Omega, \\ u_i(0) = u_{i0} & \text{in } \Omega. \end{cases} \tag{3a}$$

The effective diffusion matrix  $D_i(r) \in \mathbb{R}^{2 \times 2}$  can be calculated using any solution  $w_k$ , k = 1, 2, of the cell problem

$$(P_{ii}) \begin{cases} -\Delta w_k = 0 & \text{in } S \times (Y \setminus \overline{B}(r)), \\ -\nabla w_k \cdot \nu = e_k \cdot \nu & \text{on } S \times \partial B(r), \\ y \mapsto w_k & \text{is } Y\text{-periodic.} \end{cases} \tag{3d}$$

Here,  $Y=(0,1)^2$  denotes the unit cell,  $\overline{B}(r)$  is the closed ball with radius r and center point  $\overline{y}=(1/2,1/2)$ , and  $e_k$  the k-th unit normal vector. Finally,  $\nu=\nu(y)$  denotes the outer unit normal vector of  $Y\setminus \overline{B}(r)$ . We then have  $(d_i>0$  are known constants)

$$(D_i)_{jk} = d_i \phi(r) \int_{Y \setminus \overline{B}(r)} (\nabla w_k + e_k) \cdot e_j \, dz$$

where  $\phi(r)=\frac{1-\pi r^2}{|Y|}$  denotes the porosity of the medium. In our case, since |Y|=1, we get  $\phi(r)=1-\pi r^2$ . For more details regarding the cell problem and the effective diffusivity, we refer to [26] where they are established via homogenization.

Finally, the evolution of v is governed by an ODE parametrized in  $x \in \Omega$ 

$$(P_{iii}) \quad \begin{cases} \partial_t v = \sum_{i=1}^N \left( a_i u_i - b_i v \right) & \text{in } S \times \Omega, \\ v(0) = v_0 & \text{in } \Omega. \end{cases}$$
 (3g)

and the radius function is governed by the following ODE parametrized in  $x\in\Omega$ 

$$(P_{iv}) \quad \begin{cases} \partial_t r = 2\pi\alpha \sum_{i=1}^N \left(a_i u_i - b_i v\right) & \text{in } S \times \Omega, \\ r(0) = r_0 & \text{in } \Omega. \end{cases}$$
 (3i)

where  $\alpha > 0$  is an additional proportionality coefficient. We have labeled these subproblems  $(P_i) - (P_{iv})$  for easier referencing.

**Remark 2.** Please note that changes in time of v and r are closely connected via  $2\pi\alpha\partial_t v=\partial_t r$  or, more explicitly,

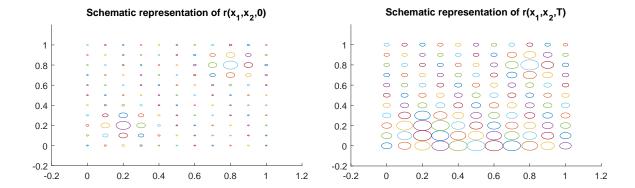
$$r(t) = r_0 - 2\pi\alpha(v(t) - v_0). \tag{4}$$

For that reason, Problem  $(P_{iv})$  (equations (3i) and (3j)) can be substituted with equation (4).

A possible initial choice for the radii  $r_0$  is depicted in Figure 1. We point out there also what will happen at the final time T; more details on the parameter setup are given in the simulation sections. What concerns the modeling of the deposition of the colloidal populations, our choice is similar to one reported in [20].

This accounts for the simple observation that the absorbed material leads to the clogging of the pore under the fundamental assumption of the growth of the radius is proportional to the amount of

<sup>&</sup>lt;sup>1</sup>Note that  $w_k$  is only unique up to a constant: this is fine since we are only interested in  $\nabla w_k$  later.



**Figure 1:** Example of  $r(x_1, x_2, t = 0)$  with corresponding  $r(x_1, x_2, t = T)$  of the same simulation. The parameter setting is as discussed in Figure 13. Regions with larger circles correspond to low porosity and permeability.

material that is absorbed. For a more concrete argumentation for this particular structure, we again point to [26].

The overall problem we are considering in this work is then given by equations (3a–3j). Regarding our concept of a weak solution of this system:

**Definiton 3** (Weak solution). For a time interval  $(0,s) \subset S$ , a weak solution to the problem is given by a set of functions (u,v,w,r) with the regularity

$$u_i \in L^2((0,s); H^1(\Omega)) \cap L^{\infty}((0,s) \times \Omega)$$
 such that  $\partial_t u_i \in L^2((0,s) \times \Omega)$ ,  $w \in L^2((0,s) \times \Omega; H^1_{\#}(Y)), v \in W^{1,1}((0,s); L^2(\Omega)), r \in W^{1,1}((0,s); L^2(\Omega))$ 

that satisfies equations (3a-3j) in the standard weak Sobolev setting.

**Solution strategy.** Without yet caring about regularity issues (like smoothness, integrability, measurability) and possible singularities, we want to suggest our solution strategy for the problem given by equations (3a–3j) and show how it relates to the abstract quasi-linear PDE System 1.

We start with a few comments regarding the particular structure of our problem where we refer to the subproblems  $(P_i) - (P_{iv})$  for u, w, v, r, viz.

- The problem is strongly coupled:  $(P_i)$  depends on  $u, w, v, r, (P_{ii})$  on  $w, r, (P_{iii})$  on u, v, and  $(P_{iv})$  on r, u, v.
- Problem  $(P_i)$  is parabolic in u,  $(P_{ii})$  elliptic in w,  $(P_{iii})$  and  $(P_{iv})$  are first order ODEs in v and r.
- Problem  $(P_i)$  is nonlinear in u and r,  $(P_{ii})$  is nonlinear in r, and  $(P_{iii})$  and  $(P_{iv})$  are linear.
- Problem  $(P_{ii})$  is not a *real* free boundary problem, as the underlying domain  $Y \setminus \overline{B(r)}$  depends only on (t,x) while the derivatives are w.r.t. y.

As a consequence of these points, a natural strategy is to first tackle the ODEs and to use them to inform the cell problem and the parabolic system. In the following, we outline the intermediate steps involved in getting to the abstract fixed-point problem that will be the starting point for our analysis in Theorem 13.

Step (a): Looking at the linear ODE for v (Problem  $(P_{iii})$ ), we find the characterization of v in terms of u via (setting  $b = \sum_{i=1}^N b_i$ )

$$v(t,x) = e^{-bt} \left( v_0(x) + \sum_{i=1}^{N} a_i \int_0^t e^{b\tau} u_i(\tau, x) d\tau \right).$$

With this in mind, we can eliminate v for u in our problem by setting  $v = \mathcal{L}_v(u)$ , where  $\mathcal{L}_v$  is the abstract solution operator for the v-problem.

Step (b): Similarly, looking at the second ODE (Problem  $(P_{iv})$ ), we have

$$r(t,x) = r_0(x) + 2\pi\alpha \sum_{i=1}^{N} \int_0^t (a_i u_i(\tau, x) - b_i v(\tau, x)) d\tau$$

With this characterization, we can introduce the corresponding solution operator  $\widetilde{\mathcal{L}_r}$  via

$$r = \mathcal{L}_r(u, v) = \mathcal{L}_r(u, \mathcal{L}_v(u)) = \widetilde{\mathcal{L}_r}(u).$$

Step (c): Looking at the cell problem (k=1,2) (Problem  $(P_{ii})$ )

$$\begin{split} -\Delta w_k &= 0 & \text{in } S \times (Y \setminus \overline{B}(r)), \\ -\nabla w_k \cdot \nu &= e_k \cdot \nu & \text{on } S \times \partial B(r), \\ y &\mapsto \tau(\cdot, \cdot, y) & \text{is } Y\text{-periodic}, \end{split}$$

we expect to get solutions for every given r>0 such that  $\overline{B}(r)\cap\partial Y=\emptyset$ . We introduce the corresponding solution operator via

$$w = \mathcal{L}_w(r) = \left(\mathcal{L}_w \circ \widetilde{\mathcal{L}_r}\right)(u) = \widetilde{\mathcal{L}_w}(u).$$

Step (d): Putting everything together, we can rewrite the parabolic problem  $(P_i)$ 

$$\partial_t u_i - \operatorname{div}(D_i(r, w)\nabla u_i) = R_i(u) - \frac{2\pi r}{1 - \pi r^2}(a_i u_i - b_i v)$$

into

$$\partial_t u_i - \operatorname{div}\left(D_i\left(\widetilde{\mathcal{L}_r}(u), \widetilde{\mathcal{L}_w}(u)\right) \nabla u_i\right) = R_i(u) - \frac{2\pi \widetilde{\mathcal{L}_r}(u)}{1 - \pi (\widetilde{\mathcal{L}_r}(u))^2} (a_i u_i - b_i \mathcal{L}_v(u))$$

This highly nonlinear system of PDEs is now given only in terms of the unknown function u. On an abstract level, we therefore want to investigate parabolic systems in the form of

$$\partial_t u_i - \operatorname{div}\left(\widehat{D}_i(u)\nabla u_i\right) = F_i(u) \quad \text{in } S \times \Omega,$$
 (5a)

$$-\widehat{D}_i(u)\nabla u_i\cdot n=0 \qquad \text{ on } S\times\partial\Omega, \tag{5b}$$

$$u_i(0) = u_{i0}$$
 in  $\Omega$  (5c)

where

$$F_i(u) = R_i(u) - \frac{2\pi \widetilde{\mathcal{L}_r}(u)}{1 - \pi (\widetilde{\mathcal{L}_r}(u))^2} (a_i u_i - b_i \mathcal{L}_v(u)).$$

The exact setting regarding function spaces will be settled in the following section.

# 3 Analysis

In this section, we present the detailed fixed-point argument (as outlined in Section 2) for the non-linear system given via equations (5a–5c). The corresponding solution u can then be used to get the solution (u, v, w, r) of the overall problem (equations (3a–3j)) via the solution operators  $\mathcal{L}_v$ ,  $\mathcal{L}_r$ , and  $\mathcal{L}_w$  as introduced in Section 2.

The strategy of our proof is a three-step process:

1) For a given function  $\tilde{u}$  (of sufficient regularity), we establish well-posedness and estimates for the linear problem given by

$$\partial_t u_i - \operatorname{div}\left(\widehat{D}_i(\tilde{u})\nabla u_i\right) = F_i(\tilde{u}) \quad \text{in } S \times \Omega,$$
 (6a)

$$-\widehat{D_i}(\tilde{u})\nabla u_i\cdot n=0 \qquad \text{on } S\times\partial\Omega, \qquad \text{(6b)}$$
 
$$u_i(0)=u_{i0} \qquad \text{in } \Omega. \qquad \text{(6c)}$$

$$u_i(0) = u_{i0} \qquad \text{in } \Omega. \tag{6c}$$

This is established in Lemma 9.

- 2) We show that there is a set of functions such that the solution operator for equations (6a–6c) maps that set onto itself, see Lemma 11. This result is local in time, since we need to keep t small in order to control the norm of the solution.
- 3) Finally, we employ Schauder's fixed point theorem to establish the existence of at least one solution, see Theorem 13.

This fixed-point argument is outlined in Section 3.2. Before that, in Section 3.1, we collect some important technical auxiliary results that are needed in the analysis.

In the following, for some later to be fixed M>0 and  $s\in(0,T)$ , let

$$T_{s,M} = \{ u \in L^2((0,s) \times \Omega)^N : ||u_i||_{\infty} \le M \ (i = 1,...,N) \}.$$

For ease of notation, for any given u of sufficient regularity we will write  $v_u = \mathcal{L}_v(u)$ ,  $r_u = \widetilde{\mathcal{L}_r}(u)$ ,  $w_u = \overline{\mathcal{L}_w}(u)$  for the corresponding solution given for the particular subproblem and  $Y_u = Y \setminus \overline{B(r_u)}$ .

### 3.1 Auxiliary results

We start by collecting some important auxiliary results and estimates that will be needed in the construction of the actual fixed-point argument.

Function	Assumption	Reason
$r_0$	$1/8 \le r_0(x) \le 1/4$	Room for growth and shrinkage
$u_{i0}$	$0 \le u_{i0}(x) \le M/2$	Keeping the solution in $T_{s,M}$
$v_0$	$0 \le v_0(x) \le C_v$	Bounding $v_u$

Table 1: Assumptions regarding the initial data.

In a first step, we establish some sufficient conditions for the diffusivity matrix to not degenerate. Note that at this point it is not clear that this condition actually can be satisfied; this is shown in Lemma 5.

**Lemma 4** (Diffusivity). If  $u \in L^2((0,s) \times \Omega)$  is chosen such that  $0 \le 2r_u \le 1 - \varepsilon_1$  for some small  $\varepsilon_1 > 0$ , we find that  $\widehat{D}_i(u)$  is symmetric and positive definite, i.e.,  $\widehat{D}_i(u)\xi \cdot \xi \geq c_i |\xi|^2$  where the constants  $c_i > 0$  do not depend on u and  $\xi \in \mathbb{R}^2$ . In addition,  $\widehat{D}_i(u) \in L^{\infty}((0,s) \times \Omega)$ .

*Proof.* This is a rather standard argument in homogenization scenarios with fixed geometry, see, e.g., [13, 17]. Due to our restrictions on the radius function r, this argument can easily be transferred to our situation. The entries of  $D_i$  are given by

$$(\widehat{D}_i(u))_{jk} = d_i \phi(r_u) \int_{Y_u} (\nabla w_{u,k} + e_k) \cdot e_j \, dz$$

where  $r_u = \widetilde{\mathcal{L}_v}(u)$ ,  $Y_u = Y \setminus \overline{B}(r_u)$ , and  $w_u = (w_{u,1}, w_{u,2}) = \widetilde{\mathcal{L}_w}(u)$ . The  $D_i$  are symmetric since

$$\int_{Y_u} (\nabla w_{u,k} + e_k) \cdot e_j \, dz = \int_{Y_u} (\nabla w_{u,k} + e_k) \cdot (\nabla w_{u,j} + e_j) \, dz$$

by way of  $w_{u,k}$  solving the cell problem.

Via that representation, non negativity is also straightforward to show (we refer to [30, Section 12.5] for a similar argument) as long as  $\phi(r_u)$  is non negative. For the positivity, we have to ensure that there is some  $c_i>0$  such that  $\phi(r_u),\,|Y_u|\geq c_i$  for all  $(t,x)\in S\times\Omega$ . Both hold true if  $r_u$  is bounded away from 1/2, i.e, if there is some  $\varepsilon_1>0$  such that  $2r_u\leq 1-\varepsilon_1$  for all  $(t,x)\in S\times\Omega$ .

Now, regarding the boundedness of  $D_i$ , we first note that  $|\phi(r_u)| \leq |Y|^{-1} = 1$  when  $0 \leq 2r_u \leq 1 - \varepsilon_1$  is satisfied. Using Hölder's inequality, we can estimate

$$\left| (\widehat{D_i}(u))_{jk} \right| \le \frac{d_i}{2} \left( 3|Y_u| + \int_{Y_u} |\nabla w_{u,k}|^2 dy \right).$$

Standard energy estimates for  $w_{u,k}$  (testing the system given by equations (3a–3c) with  $w_{k,u}$ ) lead to

$$\int_{Y_u} |\nabla w_{u,k}|^2 dy \le \frac{|Y_u|}{2} + \int_{\partial Y_u} |w_{u,k}| ds.$$

Choosing, without loss of generality, the solution  $w_{u,k}$  with zero average, we can employ the trace theorem and a version of the Poincaré inequality and estimate

$$\int_{Y} |\nabla w_{u,k}|^2 \, \mathrm{d}y \le C(r_u)|Y_u|.$$

where the constant  $C(r_u)$  (for the trace operator and Poincaré inequality) scales monotonous with the radius  $r_u$  and where  $|Y_u| \le |Y| \le 1$ . Due to our restriction  $2r_u \le 1 - \varepsilon_1$ , we can estimate

$$\left| (\widehat{D}_i(u))_{jk} \right| \leq \frac{d_i}{2} (3 + C(1)).$$

In the following, we will establish sufficient conditions for a function  $u \in L^2((0,s) \times \Omega)$  to guarantee that the condition  $2r_u \le 1 - \varepsilon_1$  is met. Setting

$$a_u(t,x) = 2\pi\alpha \sum_{i=1}^{N} (a_i u_i - b_i v_u),$$

we get

$$r_u(t,x) = r_0(x) + \int_0^t a_u(\tau, x) d\tau.$$
 (7)

**Lemma 5** (Bounds for r). Let  $M, \varepsilon_1, \varepsilon_2 > 0$  satisfy

$$M\left(e^{bt} - 1\right) \le \frac{b}{2\pi\alpha a} \min\{1 - 2\sup r_0 - \varepsilon_1, \inf r_0 - \varepsilon_2 - 2\pi\alpha bt \sup v_0\}$$
(8)

for all  $t \in (0,s)$ . Here,  $a = \sum a_i$  and  $b = \sum b_i$ . Then,  $2\varepsilon_2 \le 2r_u \le 1 - \varepsilon_1$  for all  $u \in T_{s,M}$ .

*Proof.* For every  $u \in T_{s,M}$ , we find that

$$v_u(t,x) = e^{-bt} \left( v_0(x) + \sum_{i=1}^N a_i \int_0^t e^{b\tau} u_i(\tau,x) d\tau \right).$$

As a consequence,

$$-\frac{a}{b}M(e^{bt} - 1) \le v_u(t, x) \le v_0(x) + \frac{a}{b}M(e^{bt} - 1).$$

This implies

$$a_u = 2\pi\alpha \sum_{i=1}^{N} (a_i u_i - b_i v_u) \le 2\pi\alpha \left(aM + aM(e^{bt} - 1)\right) = 2\pi\alpha aMe^{bt}$$

as well as

$$a_u \ge -2\pi\alpha \left(aM + bv_0(x) + aM(e^{bt} - 1)\right) = -2\pi\alpha \left(bv_0(x) + aMe^{bt}\right).$$

Therefore.

$$\inf r_0 - 2\pi\alpha \left(tb\sup v_0 + \frac{aM}{b}(e^{bt} - 1)\right) \le r_u(t, x) \le \sup r_0 + 2\pi \frac{\alpha aM}{b}(e^{bt} - 1).$$

As a consequence,  $2\varepsilon_2 < 2r_u < 1 - \varepsilon_1$  can be ensured by the following two conditions:

$$M\left(e^{bt} - 1\right) \le \frac{b}{2\pi\alpha a} \left(1 - 2\sup r_0 - \varepsilon_1\right),$$
  
$$M\left(e^{bt} - 1\right) \le \frac{b}{2\pi\alpha a} \left(\inf r_0 - \varepsilon_2 - 2\pi\alpha bt \sup v_0\right).$$

**Remark 6.** The condition 8 required in Lemma 5 can always be met (over some possibly small time interval (0,s)) for  $M, \varepsilon_1, \varepsilon_2$  small enough as long as the initial radius distribution satisfies  $2\varepsilon_2 < 2r_0(x) < 1 - \varepsilon_1$ . Connecting Lemma 5 with Lemma 4 leads to well behaved diffusivities for  $u \in T_{s,M}$ .

Now, looking at the r.h.s. of our reaction diffusion equation, we have for  $u \in T_{s,M}$  (setting  $\gamma = \max_{i,j} \gamma_{ij}$ ):

$$-M^{2}\gamma\left(N-\frac{k+1}{2}\right) \leq R_{k}(u) \leq M^{2}\gamma\left(N-\frac{k+1}{2}\right) \quad (1 \leq k \leq N). \tag{9}$$

Due to  $r_u \leq 1/2$  and

$$\frac{2\pi r_u}{1 - \pi r_u^2} \le \frac{\pi}{1 - \pi/4} \le 15$$

we arrive at

$$\frac{2\pi r_u}{1 - \pi r_u^2} (a_i u_i - b_i v_u) \le 15 \left( a_i M + \frac{a}{b} b_i M(e^{bt} - 1) \right), \tag{10}$$

and

$$\frac{2\pi r_u}{1 - \pi r_u^2} (a_i u_i - b_i v_u) \ge -15 \left( a_i M + b_i \left( v_0(x) + \frac{a}{b} M(e^{bt} - 1) \right) \right). \tag{11}$$

As a consequence, for every  $u \in T_{s,M}$ , we find that  $F_i(u) \in L^{\infty}(S \times \Omega)$  for all i = 1, ..., N. In particular, we find that

$$\sup\{\|F_i(u)\|_{\infty} : u \in T_{s,M}\} = C \tag{12}$$

where the constant C depends only on s, M.

**Lemma 7** (Estimates for the radius). For  $u^{(1)}, u^{(2)} \in T_{s,M}$  let  $r^{(1)}, r^{(2)}$  be the corresponding solutions of the radius ODE problem. Then,

$$\left| r^{(1)} - r^{(2)} \right| \le C \int_0^t \left( \left\| u^{(1)} - u^{(2)} \right\|_{\infty} + \int_0^{\tau} e^{bs} \left\| u^{(1)} - u^{(2)} \right\|_{\infty} ds \right) d\tau.$$

where the constant C>0 is independent of the particular choice of  $u^{(k)}$  (k=1,2)

*Proof.* The radius ODE can be solved by integration (k = 1, 2):

$$r^{(k)}(t,x) = r_0(x) + 2\pi\alpha \sum_{i=1}^{N} \int_0^t a_i u_i^{(k)}(\tau,x) - b_i v^{(k)}(\tau,x) d\tau$$

where  $v^{(k)}$  are given via

$$v^{(k)}(t,x) = e^{-bt} \left( v_0(x) + \sum_{i=1}^N a_i \int_0^t e^{b\tau} u_i^{(k)}(\tau,x) d\tau \right).$$

Consequently, we can estimate

$$\left| r^{(1)} - r^{(2)} \right| \le 2\pi\alpha \sum_{i=1}^{N} \int_{0}^{t} \left( a_{i} \left| u_{i}^{(1)} - u_{i}^{(2)} \right| + b_{i} \sum_{j=1}^{N} a_{j} \int_{0}^{\tau} e^{bs} \left| u_{j}^{(1)} - u_{j}^{(2)} \right| \, \mathrm{d}s \right) \, \mathrm{d}\tau$$

$$\le C \int_{0}^{t} \left( \left\| u^{(1)} - u^{(2)} \right\|_{\infty} + \int_{0}^{\tau} e^{bs} \left\| u^{(1)} - u^{(2)} \right\|_{\infty} \, \mathrm{d}s \right) \, \mathrm{d}\tau.$$

where the constant C>0 is independent of the particular choice of  $u^{(k)}$  (k=1,2).

**Lemma 8** (Estimates for the cell problem). Let  $\varepsilon_2 \leq r_1 \leq r_2 \leq 1/2(1-\varepsilon_1)$  and let  $w_k^{(i)}$ , k, i=1,2, solve

$$\begin{split} -\Delta w_k^{(i)} &= 0 & \text{in } S \times Y^{(i)}, \\ -\nabla w_k^{(i)} \cdot \nu &= e_k \cdot \nu & \text{on } S \times \Sigma^{(i)}, \\ \int_{Y^{(i)}} w_k^{(i)}(y) \, \mathrm{d}y &= 0, \\ y &\mapsto w_k^{(i)}(y) & \text{is $Y$-periodic.} \end{split}$$

Then, the following estimate holds:

$$\left| \int_{Y^{(1)}} \nabla w_k^{(1)} \cdot e_j \, \mathrm{d}y - \int_{Y^{(2)}} \nabla w_k^{(2)} \cdot e_j \, \mathrm{d}y \right| \le C|r^{(1)} - r^{(2)}|,$$

where the constant C>0 might depend on  $e_1$  and  $e_2$  but not on the particular choice of  $r^{(1)}$  and  $r^{(2)}$ . Here, we have set  $Y^{(j)}=Y\setminus \overline{B(r^{(j)})}$  and  $\Sigma^{(j)}=\partial B(r^{(j)})$ , where  $B(r^{(j)})$  is the closed ball with radius  $r^{(j)}$  and center point  $\overline{y}=(1/2,1/2)$ .

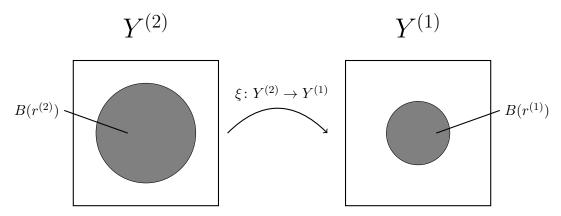
*Proof.* We prove this statement in three steps. First, we introduce a coordinate transform that allows us to compare the different solutions and, second, go on proving some important energy estimates. Finally, we use these energy estimates to proof the desired result.

Step1: Transformation: We set  $\overline{y}=(1/2,1/2)$  and introduce the transformation  $\xi\colon\overline{Y}\to\overline{Y}$  given by

$$\xi(y) = \begin{cases} y, & |y - \overline{y}| \ge 1/2, \\ (1 - \chi(|y - \overline{y}|))y + \chi(|y - \overline{y}|) \left(r^{(1)}/r^{(2)}(y - \overline{y}) + \overline{y}\right), & r^{(2)} \le |y - \overline{y}| \le 1/2, \\ r^{(1)}/r^{(2)}(y - \overline{y}) + \overline{y}, & |y - \overline{y}| \le r^{(2)}. \end{cases}$$

Here,  $\chi\colon [r^{(2)},{}^1\!/2]\to [0,1]$  is a smooth cut-off function with compact support (i.e.,  $\chi\in C_0^\infty(r^{(2)},{}^1\!/2)$ ) satisfying  $\chi(r^{(2)})=1$ ,  $\chi({}^1\!/2)=0$ , as well as  $-4/\varepsilon_1\leq \chi'(z)\leq 0$ .

Regarding this transformation: in the area where  $|y-\overline{y}| \leq r^{(2)}$ , it is a scaling which transforms the larger ball with radius  $r^{(2)}$  into the smaller ball with radius  $r^{(1)}$  (implying  $\xi(Y^{(2)}) = Y^{(1)}$ ). In the area where  $|y-\overline{y}| \geq 1/2$ , it is the identity function (implying  $\nu_{\Sigma^{(1)}}(\xi(y)) = \nu_{\Sigma^{(2)}}(y)$  for all  $y \in \Sigma^{(2)}$ ). Finally, in the transition area characterized by  $r^{(2)} \leq |y-\overline{y}| \leq 1/2$ , the cut-off function  $\chi$  is used to ensure smoothness of the transformation via interpolation.



**Figure 2:** Sketch of the transformation connecting reference cells for different radii  $r^{(1)}$  and  $r^{(2)}$ .

Calculating the Jacobi matrix for  $\xi$ , we see that  $D\xi=\mathbb{I}_2$  for  $|y-\overline{y}|\geq 1/2$  (implying  $\det D\xi=1$ ) and  $D\xi=\left(r^{(1)}/r^{(2)}\right)^2\mathbb{I}_2$  for  $|y-\overline{y}|\leq r^{(2)}$  (implying  $\det D\xi=\left(r^{(1)}/r^{(2)}\right)^2$ ). For the transition part, i.e.,  $r^{(2)}\leq |y-\overline{y}|\leq 1/2$ , we calculate

$$\begin{split} \partial_{y_i} \xi_j(y) &= \partial_{y_i} \left[ y \mapsto (1 - \chi(|y - \overline{y}|)) y_j + \chi(|y - \overline{y}|) \left( r^{(1)} / r^{(2)} (y_j - 1/2) + 1/2 \right) \right] \\ &= \delta_{ij} \left( 1 + (r^{(1)} / r^{(2)} - 1) \chi(|y - \overline{y}|) \right) + \left( r^{(1)} / r^{(2)} (y_j - 1/2) + 1/2 - y_j \right) \frac{y_i - 1/2}{|y - \overline{y}|} \chi'(|y - \overline{y}|) \end{split}$$

As a consequence, we find that the Jacobian is given by the symmetric matrix

$$D\xi(y) = a(|y - \overline{y}|) \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + b(|y - \overline{y}|) \begin{pmatrix} (y_1 - 1/2)^2 & (y_1 - 1/2)(y_2 - 1/2) \\ (y_1 - 1/2)(y_2 - 1/2) & (y_2 - 1/2)^2 \end{pmatrix}$$
(13)

where (setting  $\overline{r} = r^{(2)} - r^{(1)} \ge 0$ )

$$a(z) = \left(1 - \frac{\overline{r}}{r^{(2)}}\chi(z)\right), \quad b(z) = -\frac{\chi'(z)}{z}\frac{\overline{r}}{r^{(2)}}.$$

Therefore, this symmetric matrix takes the abstract form ( $c_i = y_i - 1/2$ )

$$D\xi = a \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + b \begin{pmatrix} c_1^2 & c_1c_2 \\ c_1c_2 & c_2^2 \end{pmatrix}$$

We calculate the corresponding determinant as

$$\det D\xi = a \Big[ a + b(c_1^2 + c_2^2) \Big].$$

or, more explicitly,

$$\det D\xi(y) = a(|y - \overline{y}|) \left[ a(|y - \overline{y}|) + b(|y - \overline{y}|) \left( y_1^2 - y_1 + y_2^2 - y_2 + 1 \right) \right].$$

Since  $a(|y-\overline{y}|) > 0$ ,  $b(|y-\overline{y}|) \ge 0$  and  $y_1^2 - y_1 + y_2^2 - y_2 + 1 > 0$  for all  $y = (y_1, y_2) \in Y$ , we find that

$$\det D\xi(y) \ge \inf_{r^{(2)} \le |y - \overline{y}| \le 1/2} a^2(|y - \overline{y}|) = \left(\frac{r^{(1)}}{r^{(2)}}\right)^2.$$

This shows that (since  $\det D\xi=1$  for  $|y-\overline{y}|\geq 1/2$  and  $\det D\xi=\left(r^{(1)}/r^{(2)}\right)^2$  for  $|y-\overline{y}|\leq r^{(2)}$ )

$$4\varepsilon_2^2 \le \left(\frac{\varepsilon_2}{1/2(1-\varepsilon_1)}\right)^2 \le \det D\xi(y) \le 1$$

which implies invertibility of  $D\xi$ .

Step 2: Energy estimates. In the following, we set  $F(y) = D\xi(y)$  and  $J(y) = |\det F(y)|$ . We start with the weak forms

$$\int_{Y^{(i)}} \nabla w_k^{(i)} \cdot \nabla \eta^{(i)} \, \mathrm{d}z = \int_{\Sigma^{(i)}} e_k \cdot \nu_{\Sigma^{(i)}} \eta^{(i)} \, \mathrm{d}\sigma$$

with text functions  $\eta^{(i)} \in H^1_\#(Y^{(i)})$  and i=1,2. We take the difference of these two weak forms:

$$\int_{Y^{(1)}} \nabla w_k^{(1)} \cdot \nabla \eta^{(1)} \, \mathrm{d}y - \int_{Y^{(2)}} \nabla w_k^{(2)} \cdot \nabla \eta^{(2)} \, \mathrm{d}y = e_k \cdot \left[ \int_{\Sigma^{(1)}} \nu_{\Sigma^{(1)}} \eta^{(1)} \, \mathrm{d}\sigma - \int_{\Sigma^{(2)}} \nu_{\Sigma^{(2)}} \eta^{(2)} \, \mathrm{d}\sigma \right].$$

and transform the surface integral on the right-hand side in order to arrive at

$$\int_{\Sigma^{(1)}} \nu_{\Sigma^{(1)}} \eta^{(1)} d\sigma - \int_{\Sigma^{(2)}} \nu_{\Sigma^{(2)}} \eta^{(2)} d\sigma = \int_{\Sigma^{(2)}} \nu_{\Sigma^{(1)}} (\xi(y)) \eta^{(1)} (\xi(y)) J(y) d\sigma - \int_{\Sigma^{(2)}} \nu_{\Sigma^{(2)}} \eta^{(2)} d\sigma.$$

By construction, we have  $\nu_{\Sigma^{(1)}}(\xi(y))=\nu_{\Sigma^{(2)}}(y)$  for all  $y\in\Sigma^{(2)}$  leading to

$$\begin{split} \int_{\Sigma^{(1)}} \nu_{\Sigma^{(1)}} \eta^{(1)} \, \mathrm{d}\sigma - \int_{\Sigma^{(2)}} \nu_{\Sigma^{(2)}} \eta^{(2)} \, \mathrm{d}\sigma &= \int_{\Sigma^{(2)}} \left( \eta^{(1)}(\xi(y)) J(y) - \eta^{(2)}(y) \right) \nu_{\Sigma^{(2)}} \, \mathrm{d}\sigma \\ &= \int_{\Sigma^{(2)}} \left( \eta^{(1)}(\xi(y)) - \eta^{(2)}(y) \right) J(y) \nu_{\Sigma^{(2)}} \, \mathrm{d}\sigma \\ &+ \int_{\Sigma^{(2)}} \left( J(y) - 1 \right) \eta^{(2)}(y) \nu_{\Sigma^{(2)}} \, \mathrm{d}\sigma \end{split}$$

For the volume integral on the l.h.s., we get (note that the Jacobian is symmetric)

$$\int_{Y^{(1)}} \nabla w_k^{(1)} \cdot \nabla \eta^{(1)} \, \mathrm{d}y - \int_{Y^{(2)}} \nabla w_k^{(2)} \cdot \nabla \eta^{(2)} \, \mathrm{d}y \\
= \int_{Y^{(2)}} J(y) F(y)^{-2} \nabla w_k^{(1)}(\xi) \cdot \nabla \eta^{(1)}(\xi) - \nabla w_k^{(2)} \cdot \nabla \eta^{(2)} \, \mathrm{d}y$$

and, as a consequence,

$$\begin{split} \int_{Y^{(2)}} J(y) F(y)^{-2} \nabla w_k^{(1)}(\xi) \cdot \nabla \eta^{(1)}(\xi) - \nabla w_k^{(2)} \cdot \nabla \eta^{(2)} \, \mathrm{d}y \\ &= \int_{\Sigma^{(2)}} \bigg( \eta^{(1)}(\xi(y)) - \eta^{(2)}(y) \bigg) J(y) \nu_{\Sigma^{(2)}} \, \mathrm{d}\sigma + \int_{\Sigma^{(2)}} \bigg( J(y) - 1 \bigg) \eta^{(2)}(y) \nu_{\Sigma^{(2)}} \, \mathrm{d}\sigma. \end{split}$$

Choosing text functions  $\eta^{(1)}(\xi) = \eta^{(2)} = w_k^{(1)}(\xi) - w_k^{(2)} =: \overline{w}_k$ , this leads to

$$\|\nabla \overline{w}_k\|_{L^2(Y^{(2)})}^2 \le \int_{Y^{(2)}} |J(y)F(y)^{-2} - \mathbb{I}_2| |\nabla w_k^{(1)}(\xi)| \cdot |\nabla \overline{w}_k| \, dy + \int_{\Sigma^{(2)}} |J(y) - 1| |\overline{w}_k| \, d\sigma.$$

For  $y\in \Sigma^{(2)}$ , i.e.,  $|y-\overline{y}|=r^{(2)}$ , we have

$$1 - J(y) = 1 - \left(\frac{r^{(1)}}{r^{(2)}}\right)^2 = \frac{(r^{(2)})^2 - (r^{(1)})^2}{(r^{(2)})^2} \le \frac{\overline{r}}{r^{(2)}}.$$

Now, for  $y\in Y_2$  with  $|y-\overline{y}|\geq 1/2$ , we have J(y)=1 and  $F(y)=\mathbb{I}_2$  and, in the case that  $r^{(2)}\leq |y-\overline{y}|\leq 1/2$ ,

$$|J(y)F(y)^{-2} - \mathbb{I}_2| \le \frac{|J(y) - 1|}{|F(y)|^2} + \frac{|(D\xi)^{-1} - \mathbb{I}_2|}{|F(y)|} + |F(y)^{-1} - \mathbb{I}_2|$$

Since  $|F(y)|^2 \geq J(y) \geq 4\varepsilon_2^2$  and  $1-J(y) \leq \overline{r}/r^{(2)}$ :

$$|J(y)F(y)^{-2} - \mathbb{I}_2| \le \frac{\overline{r}}{4r^{(2)}\varepsilon_2^2} + \left(1 + \frac{1}{2\varepsilon_2}\right)|F(y)^{-1} - \mathbb{I}_2|$$

Finally, via

$$|F(y)^{-1} - \mathbb{I}_2| \le |F(y)^{-1}| |\mathbb{I}_2 - F(y)| \le 2\varepsilon_2 |\mathbb{I}_2 - F(y)|$$

we arrive at (looking at equation (13))

$$\left|J(y)F(y)^{-2} - \mathbb{I}_2\right| \le \frac{\overline{r}}{r^{(2)}} \left(\frac{1}{4\varepsilon_2^2} + 2\varepsilon_2 + 1 + \frac{1}{\varepsilon_1 r^{(2)}}\right)$$

Therefore we find that (note that  $r^{(2)} \geq \varepsilon_2$ )

$$\|\nabla \overline{w}_k\|_{L^2(Y^{(2)})}^2 \le C(\varepsilon_1, \varepsilon_2) \overline{r} \left( \int_{Y^{(2)}} \left| \nabla w_k^{(1)}(\xi) \right| \cdot |\nabla \overline{w}_k| \, dy + \int_{\Sigma^{(2)}} |\overline{w}_k| \, d\sigma \right).$$

Applying Poincaré's inequality (possible due to the zero average condition) and the trace theorem leads to the energy estimate

$$\|\overline{w}_k\|_{H^1(Y^{(2)})} \le \tilde{C}(\varepsilon_1, \varepsilon_2)\overline{r},$$
 (14)

where the constant  $\tilde{C}(\varepsilon_1,\varepsilon_2)>0$  is independent of the specific choices of  $r^{(1)}$  and  $r^{(2)}$ .

Step 3: Proving the result. Using equation (14), we go on by estimating the following key expression:

$$\left| \int_{Y^{(1)}} \nabla w_k^{(1)} \cdot e_j \, dy - \int_{Y^{(2)}} \nabla w_k^{(2)} \cdot e_j \, dy \right| \le \left\| \int_{Y^{(1)}} \nabla w_k^{(1)} \, dy - \int_{Y^{(2)}} \nabla w_k^{(2)} \, dy \right\|$$

$$= \left\| \int_{Y^{(2)}} J(y) F(y)^{-1} \nabla w_k^{(1)}(\xi) - \nabla w_k^{(2)} \, dy \right\|$$

$$\le \widehat{C}(\varepsilon_1, \varepsilon_2) \overline{r},$$

### 3.2 A fixed-point argument

Now, let  $\varepsilon_1, \varepsilon_2, M^*, s^* > 0$  and initial conditions  $r_0, v_0$  be chosen such that  $2\varepsilon_2 \leq 2r_u(t,x) \leq 1 - \varepsilon_1$  for all  $(t,x) \in (0,s^*) \times \Omega$  and all  $u \in T_{s^*,M^*}$  (this is possible due to Lemmas 4 and 5). Also, let  $0 \leq u_{i0}(x) \leq M^*/2$ . These choices imply  $F(u) = (F_1(u),...,F_N(u)) \in L^\infty((0,s^*) \times \Omega)^N$  for all  $u \in T_{s^*,M^*}$  (see equation (12)). In the following, let  $s \in (0,s^*)$  and  $M \in (0,M^*)$ . We introduce the space

$$W((0,s);H^1(\Omega)) := \left\{ u \in L^2(S;H^1(\Omega)) : \partial_t u \in L^2(S;H^1(\Omega)^*) \right\}.$$

**Lemma 9** (Existence result for linearized problem). For each  $\tilde{u} \in T_{s,M}$ , there is a unique  $u \in W((0,s);H^1(\Omega))$  solving the problem given by equations (6a–6c). Moreover, the following a priori estimates are satisfied

$$\|\partial_t u\|_{L^2((0,s);H^1(\Omega)^*)}^2 + \|u\|_{L^{\infty}((0,s);L^2(\Omega))}^2 + \|\nabla u\|_{L^2((0,s)\times\Omega)}^2$$

$$\leq C \left( \|u_0\|_{L^2(\Omega)}^2 + \sum_{i=1}^N \|F_i(\tilde{u})\|_{L^2((0,s)\times\Omega)}^2 \right)$$

where the constant C>0 does not depend on  $\tilde{u}, s$ , and M. Please note that the above estimate implies boundedness in  $W((0,s);H^1(\Omega))$  as well.

*Proof.* Since  $\tilde{u} \in T_{s,M}$ , we have  $F_i(\tilde{u}) \in L^\infty((0,s) \times \Omega)$  (i=1,...,N). Also, the diffusivity matrix  $\widehat{D_i}(\tilde{u})$  is uniformly positive definite (i.e., there is  $c_i > 0$  such that  $\widehat{D_i}(\tilde{u})(t,x)\xi \cdot \xi \geq c_i |\xi|^2$  for all  $(t,x) \in (0,s) \times \Omega$  and all  $\xi \in \mathbb{R}^3$ ). Finally, as the  $D_i$  are also bounded, the existence of a unique solution follows by standard theory for linear, parabolic PDE (see, e.g., [40, Chapter 1]).

To search for the needed a priori estimates, we test the weak form with  $u_i$ . Hence, we are led to

$$||u_i(t)||_{L^2(\Omega)}^2 + 2c_i \int_0^t ||\nabla u_i||_{L^2(\Omega)}^2 d\tau \le ||u_{i0}||_{L^2(\Omega)}^2 + 2\int_0^t \int_{\Omega} |F_i(\tilde{u})u_i| dx d\tau \quad (t \in (0, s)).$$

From here, summing over i=1,...,N and applying Grönwall's inequality leads to the desired estimate for u and  $\nabla u$ . Similarly, taking a test function  $\varphi\in L^2((0,s);H^1(\Omega))$  such that  $\|\varphi\|\leq 1$ , we find that

$$\langle \partial_t u_i, \varphi \rangle_{L^2((0,s);H^1(\Omega)^*)} \le \int_{\Omega} |F_i(\tilde{u})\varphi| \,\mathrm{d}x + \int_{\Omega} |\widehat{D_i}(\tilde{u})\nabla u_i \nabla \varphi| \,\mathrm{d}x$$

thus completing the estimate.

With the solvability of the linarized problem established, we want to investigate under what circumstances we can ensure that  $u \in T_{s,M}$  as well; as this would then naturally lead to a potential

fixed-point scheme. As a first point, any  $\widetilde{u} \in T_{s,M}$  leads to a solution  $u \in W((0,s);H^1(\Omega))$  which again leads to the corresponding solution operator

$$\mathcal{L}: T_{s,M} \to W((0,s); H^1(\Omega))^N$$
.

We now need to show, that  $s \in (0, s^*)$  and  $M \in (0, M^*)$  can be chosen such that  $\mathcal{L}[T_{s,M}] \subset T_{s,M}$ . With the following lemma, we first establish  $\mathcal{L}[T_{s,M}] \subset L^{\infty}((0,s) \times \Omega)^N$ .

**Lemma 10** (Boundedness). For every  $\tilde{u} \in T_{s,M}$ , the solution of the linearized equation is bounded by

$$-t \operatorname{ess\,sup}(F_i(\tilde{u}))_- \le u_i \le \operatorname{ess\,sup} u_{i0} + t \operatorname{ess\,sup} F_i(\tilde{u}).$$

In particular, we have  $u \in L^{\infty}((0,s) \times \Omega)^N$ .

*Proof.* By the linearity of the problem, we can decompose the solution  $u_i = \pi_i + \omega_i$ , where

$$\begin{split} \partial_t \pi_i - \operatorname{div} \left( \widehat{D_i}(\tilde{u}) \nabla \pi_i \right) &= 0 & \text{ in } S \times \Omega, \\ -\widehat{D_i}(\tilde{u}) \nabla \pi_i \cdot n &= 0 & \text{ on } S \times \partial \Omega, \\ \pi_i(0) &= u_{i0} & \text{ in } \Omega, \end{split} \qquad \begin{array}{l} \partial_t \omega_i - \operatorname{div} \left( \widehat{D_i}(\tilde{u}) \nabla \omega_i \right) &= F_i(\tilde{u}) & \text{ in } S \times \Omega, \\ -\widehat{D_i}(\tilde{u}) \nabla \omega_i \cdot n &= 0 & \text{ on } S \times \partial \Omega, \\ \omega_i(0) &= 0 & \text{ in } \Omega. \end{split}$$

Estimating the  $\pi_i$ -problems via  $(\pi_i - L_i)_+$  for  $L_i = \operatorname{ess\,sup} u_{i0}$ , we find that  $\pi_i \leq L_i$ . Using Duhamel's principle ([33, Chapter 5]), we get  $\omega_i(t,x) = \int_0^t h_i(\tau,t,x) \,\mathrm{d}\tau$  where the  $\tau$ -parametrized function  $h_i$  solves

$$\begin{split} \partial_t h_i - \operatorname{div}\left(\widehat{D}_i(\tilde{u}) \nabla h_i\right) &= 0 & \text{in } S \times \Omega, \\ -\widehat{D}_i(\tilde{u}) \nabla h_i \cdot n &= 0 & \text{on } S \times \partial \Omega, \\ h_i(0) &= F_i(\tilde{u}(\tau, \cdot)) & \text{in } \Omega. \end{split}$$

This implies  $h_i \leq \operatorname{ess\,sup}(F_i(\tilde{u}))_+$  and, as a consequence  $\omega_i \leq t \operatorname{ess\,sup}(F_i(\tilde{u}))_+$ . Finally, we have

$$u_i \leq \operatorname{ess\,sup} u_{i0} + t \operatorname{ess\,sup}(F_i(\tilde{u}))_+.$$

Now, since  $u_{i0} \geq 0$ , we find that  $\pi_i \geq 0$  as well. Testing with  $(h_i + \operatorname{ess\,sup}(F_i(\tilde{u}))_-)_-$ , we arrive at  $h_i \geq -\operatorname{ess\,sup}(F_i(\tilde{u}))_-$  and, as a consequence  $\omega_i \geq -t\operatorname{ess\,sup}(F_i(\tilde{u}))_-$ . This shows

$$u_i \ge -t \operatorname{ess\,sup}(F_i(\tilde{u}))_-.$$

In particular, we find that  $u_i \in L^{\infty}((0,s) \times \Omega)$  with

$$||u_i(t)||_{L^{\infty}(\Omega)} \le ||u_{i0}||_{L^{\infty}(\Omega)} + t||F_i(\tilde{u})(t)||_{L^{\infty}(\Omega)}.$$

Now, in order to get concrete bounds for the solution  $u=(u_1,...,u_N)$ , we have to take a closer look at the right-hand sides: For the  $F_i(\tilde{u})$ , we have the estimates (given our assumptions on  $r_0$ ,  $s^*$ , and  $M^*$  and using equations (9–11)):

$$F_i(\tilde{u}) \le M\left(M\gamma\left(N - \frac{i+1}{2}\right) + 15\left(a_i + \frac{a}{b}b_i(e^{bt} - 1)\right)\right) + 15b_iv_0(x),$$

$$F_i(\tilde{u}) \ge -M\left(M\gamma\left(N - \frac{i+1}{2}\right) + 15\left(a_i + \frac{a}{b}b_i(e^{bt} - 1)\right)\right)$$

or, more compactly,

$$||F_i(\tilde{u})(t)||_{L^{\infty}(\Omega)} \le 15b_i||v_0||_{L^{\infty}(\Omega)} + M\left(M\gamma\left(N - \frac{i+1}{2}\right) + 15\left(a_i + \frac{a}{b}b_i(e^{bt} - 1)\right)\right). \tag{15}$$

With this estimate at hand, we are now able to establish that  $\mathcal{L}$  is a self-mapping for a suitable choice of (s, M).

**Lemma 11** (Self mapping property). For any  $M \in (0, M^*)$  there is  $s \in (0, s^*)$  such that for every  $\tilde{u} \in T_{s,M}$  the solution u of the linearized problem also satisfies  $u = \mathcal{L}(\tilde{u}) \in T_{s,M}$ .

*Proof.* For any given  $M \in (0, M^*)$ , we find that

$$\lim_{t \to 0} t \|F_i(\tilde{u})\|_{\infty} \to 0 \quad (i = 1, ..., N).$$

uniformly for  $\tilde{u} \in T_{s,M}$  (see inequality 15). As a consequence, it is possible to find  $s \in (0,s^*)$  such that  $s\|F_i(\tilde{u})\|_{\infty} \leq M/2$  for all i=1,...,N and for all  $\tilde{u} \in T_{s,M}$ . This implies  $u \in T_{s,M}$  via Lemma 10.

Now, let the operator  $\mathcal{L}\colon T_{s,M}\to L^2((0,s)\times\Omega)$  be the solution operator assigning any  $\tilde{u}\in T_{s,M}$  the corresponding solution of problem  $u=\mathcal{L}(\tilde{u})$  of the problem given by equations (6a–6c) (see Lemma 9). lemma 11 establishes the self mapping property of  $\mathcal{L}$ , i.e.,  $\mathcal{L}[T_{s,M}]\subset T_{s,M}$  for suitable choices of s and M. Please note that  $T_{s,M}$  is a closed subset of  $L^2((0,s)\times\Omega)$ . In the following lemma we investigate the continuity of this operator.

Lemma 12 (Continuity). The operator

$$\mathcal{L}\colon T_{s,M}\to L^2((0,s)\times\Omega)$$

is continuous with respect to the  $L^2$ -norm.

*Proof.* Now let  $\tilde{u}, \tilde{u}^{(k)} \in T_{s,M}$  such that  $\tilde{u}^{(k)} \to \tilde{u}$  in  $L^2((0,s) \times \Omega)$  for  $k \to \infty$ . In addition, let  $u = \mathcal{L}(\tilde{u})$  and  $u^{(k)} = \mathcal{L}(\tilde{u}^{(k)})$   $(k \in \mathbb{N})$  be the corresponding unique solutions to the linearized problem (see Lemma 9).

Now, the sequence  $u^{(k)}$  is bounded in  $W((0,s);H^1(\Omega))$  since  $0 \leq \tilde{u}^{(k)} \leq M$  and the a priori estimates given by Lemma 9. Since  $W((0,s);H^1(\Omega))$  is a reflexive Banach space and since it is compactly embedded in  $L^2((0,s)\times\Omega)$  (Lions-Aubin lemma), there is a subsequence (for ease of notation, still denoted by  $u^{(k)}$ ) and a limit function  $u^*$  such that  $u^{(k)}$  converges to  $u^*$  strongly and weakly in  $L^2((0,s)\times\Omega)$  and  $W((0,s);H^1(\Omega))$ , respectively. Without loss of generality, we also have  $u^{(k)}\to u$  pointwise almost everywhere over  $(0,s)\times\Omega$  (possibly by choosing a further subsequence). In the following, we show continuity by establishing that  $u^*=u.^2$ 

The components of  $u^{(k)}$  satisfy (for all  $\varphi \in H^1(\Omega)$  and  $t \in (0,s)$ )

$$\langle \partial_t u_i^{(k)}, \varphi \rangle_{H^1(\Omega)^*} + \int_{\Omega} \widehat{D}_i(\tilde{u}^{(k)}) \nabla u^{(k)} \cdot \nabla \varphi \, \mathrm{d}x = \int_{\Omega} F_i(\tilde{u}^{(k)}) \varphi \, \mathrm{d}x.$$

Now, since  $\tilde{u}^{(k)} \to \tilde{u}$  in  $L^2((0,s) \times \Omega)$ ), it holds

$$\int_{\Omega} F_i(\tilde{u}^{(k)}) \varphi \, \mathrm{d}x \to \int_{\Omega} F_i(\tilde{u}) \varphi \, \mathrm{d}x \quad (\varphi \in H^1(\Omega), \ i = 1, ..., N).$$

For the diffusion term, we take a look at

$$\int_{\Omega} \left( \widehat{D}_{i}(\tilde{u}) \nabla u^{*} - \widehat{D}_{i}(\tilde{u}^{(k)}) \nabla u^{(k)} \right) \cdot \nabla \varphi \, dx$$

$$= \int_{\Omega} \widehat{D}_{i}(\tilde{u}) \nabla \left( u^{*} - u^{(k)} \right) \cdot \nabla \varphi \, dx + \int_{\Omega} \left( \widehat{D}_{i}(\tilde{u}) - \widehat{D}_{i}(\tilde{u}^{(k)}) \right) \nabla u^{(k)} \cdot \nabla \varphi \, dx.$$

 $<sup>^2</sup>$ Due to this resulting statement: Every subsequence has a further subsequence converging to u.

Here, the first term on the right hand side goes to zero due to the weak convergence of  $u^{(k)}$  to  $u^*$  in  $W((0,s);H^1(\Omega))$ . Looking at the second term, we recall

$$\left(\widehat{D}_{i}(\widetilde{u}) - \widehat{D}_{i}(\widetilde{u}^{(k)})\right)_{lm} = d_{i} \left(\phi(r^{(0)}) \int_{Y^{(0)}} (\nabla w_{l}^{(0)} + e_{l}) \cdot e_{m} \, \mathrm{d}z - \phi(r^{(k)}) \int_{Y^{(k)}} (\nabla w_{l}^{(k)} + e_{l}) \cdot e_{m} \, \mathrm{d}z\right),$$

which can be estimated using Lemmas 7 and 8

$$\left|\widehat{D}_i(\tilde{u}) - \widehat{D}_i(\tilde{u}^{(k)})\right| \le C \int_0^t \left(\left|\tilde{u} - \tilde{u}^{(k)}\right| + \int_0^\tau e^{bs} \left|\tilde{u} - \tilde{u}^{(k)}\right| \, \mathrm{d}s\right) \, \mathrm{d}\tau.$$

Here, we have used for the porosity that

$$\left| \phi(r^{(0)}) - \phi(r^{(k)}) \right| \le \frac{\pi^2}{|Y|} \left| r^{(0)} - r^{(k)} \right|.$$

Now, since  $\tilde{u}^{(k)} \to \tilde{u}$  almost everywhere over  $(0,s) \times \Omega$ , dominated convergence leads to

$$\int_{\Omega} \left( \widehat{D}_i(\tilde{u}) - \widehat{D}_i(\tilde{u}^{(k)}) \right) \nabla u^{(k)} \cdot \nabla \varphi \, \mathrm{d}x \to 0$$

As a consequence,  $u^* = u$ .

Theorem 13 (Existence). The operator

$$\mathcal{L}: T_{s,M} \to L^2((0,s) \times \Omega)$$

has at least one fixed-point  $u^* \in W((0,s); H^1(\Omega))$ .

*Proof.*  $T_{s,M}$  is a non-empty, closed, and convex subset of  $L^2((0,s)\times\Omega)$  and  $\mathcal L$  is continuous with respect to the  $L^2((0,s)\times\Omega)$  norm (Lemma 12). Moreover, we have  $\mathcal L[T_{s,M}]\subset T_{s,M}$  via Lemma 11. Finally, since  $\mathcal L[T_{s,M}]\subset W((0,s);H^1(\Omega))$  which is compactly embedded in  $L^2((0,s)\times\Omega)$  by virtue of Lions-Aubin's lemma, we can employ Schauder's fixed point theorem to conclude the existence of at least one fixed-point  $u^*\in W((0,s);H^1(\Omega))\cap T_{s,M}$ .

**Remark 14** (Uniqueness). In general, Schauder's fixed point theorem does not lead to uniqueness of solutions. Uniqueness can, however, sometimes be established in an a posteriori way using the Lipschitz continuity of the non-linearities, when available. In our specific case, such estimates would rely on certain additional regularity assumptions that can not be met. Now, assume for a moment that we are given two solutions  $u^{(1)}, u^{(2)}$ . Standard energy estimates for the difference  $\overline{u} = u^{(1)} - u^{(2)}$  will involve the term

$$\int_{\Omega} \left( D_i(u^{(1)}) - D_i(u^{(2)}) \right) \nabla u_i^{(2)} \cdot \nabla \overline{u}_i \, \mathrm{d}x.$$

If we were able to estimate this term against

$$C \int_{\Omega} |\overline{u}| |\nabla \overline{u}| \, \mathrm{d}x,$$

the uniqueness for our problem would follow. However, such an estimate is not reachable in our case. Even if  $\nabla u^{(2)}$  were essentially bounded, the main issue is that  $u^{(1)}(t,\cdot)=u^{(2)}(t,\cdot)$  does not imply  $D_i(u^{(1)})(t,\cdot)=D_i(u^{(2)})(t,\cdot)$  because of hysteresis effects.

**Remark 15.** Relying for instance on techniques from [11], we expect the weak solution given by Theorem 13 to be of higher regularity provided that data (boundary of  $\Omega$ , initial conditions) are sufficiently smooth. This could change, however, if we were to allow actual clogging of the porous medium.

The solution as given via Theorem 13 exist only on a time interval (0,s) where s might potentially be very small (since s must be chosen to ensure the self mapping property of  $\mathcal{L}$ , see Lemma 11). The main problem in the analysis is the potential clogging as can be seen with Lemma 11 and condition (8). With that in mind, as long as it is guaranteed that clogging can not occur, i.e., r is uniformly positive and bounded away from 1/2, the solution can be extended in time indefinitely. In general, however, this will not be the case and additional careful analysis of the problem would be necessary.

# 4 Numerical simulation of the two-scale quasilinear problem

### 4.1 Setup of the model equations and target geometry

The aim is to solve numerically the two-dimensional macroscopic model problem for the species concentration  $u_i$  ( $i \in \{1, ..., N\}$ ) and v.

In the current work, we present a first attempt to treat numerically the problem by a simple combination of finite element method, for the cell problem, and a forward finite differences scheme for the macroscopic equations as we just want to observe the basic qualitative behaviour of the model towards clogging in a two dimensional (square) macroscopic domain. More advanced and accurate methods by using possibly an semi-implicit finite element scheme for the macroscopic equations, applicable as well for macroscopic domains of different shapes, should be implemented when more accuracy is desired (e.g. see [25], [28], [42]). Also, an appropriate theoretical analysis for such schemes could be additional subject of a future work.

To focus the attention on physically relevant choices of parameters, we use the setup described in [20]; see also [22, 26] for more details. Essentially, we look at a theoretical model describing the dynamics of colloid deposition on collector surfaces, when both inter-particle, and particle-surface electrostatic interactions are assumed to be negligible. The numerical range of the used parameters fit to the situations that can relate to the immobilization of bio-colloids in soils.

The simulation output we are looking after includes approximated space and time concentration profiles of colloidal populations, spatial distribution of microstructures for given time slices, and estimated amount of deposited colloidal mass. This information helps us detect in *a posteriori* way the locations in  $\Omega$  where deposition-induced clogging is likely to happen.

Initially, for readers convenience, we summarize the system of equations that we treat numerically, imposing also appropriate initial and boundary conditions for the experiments that will follow. We have

$$\partial_t u_i(t,x) = \operatorname{div}\left(D_i(r(t,x))\nabla u_i(x,t)\right) + R_i(u) - \frac{L(t,x)}{A(t,x)}\left(a_i u_i(t,x) - b_i v(t,x)\right),\tag{16a}$$

describing the diffusion of  $u_i$  in the macroscopic domain  $\Omega$ .

The effective diffusion tensor has the form

$$(D_i)_{jk}(t,x) = d_i\phi(r)\tau_{jk}(t,x),$$

where the entries

$$au_{jk} = \int_{Y \setminus \bar{B}(r)} (\nabla w_k + e_k) \, dz,$$

for all i = 1, ..., N, j, k = 1, 2.

In addition, the length L and area A functions related to the motion of the boundary (for r<1/2) are:

$$L(t,x) = \int_{\Gamma(t,x)} ds = 2\pi r(t,x), \quad A(t,x) = \int_{Y_0(t,x)} dy = 1 - \pi r^2(t,x), \quad \text{(in 2D)}$$

$$R_i(u) = \frac{1}{2} \sum_{j+l=i} \gamma_{jl} u_j u_l - u_i \sum_{j=1}^{N-i} \gamma_{ij} u_j.$$
 (16c)

Moreover, the cell functions  $w := (w_1(t, x, y), w_2(t, x, y))$ , assumed to have constant mean, in the domain Y satisfy

$$-\Delta w_k = 0, \quad i = 1, 2 \quad \text{in } Y \setminus \overline{B}(r), \tag{16d}$$

$$-\nabla w_k \cdot \nu = 0$$
, on  $\partial Y$ ,  $-\nabla w_k \cdot \nu = e_k \cdot \nu$  on  $\partial B(r)$ . (16e)

with  $\partial Y$  being the boundary of the square cell and  $n=n(t,x)=(n_1(t,x),n_2(t,x))$ , is the corresponding normal vector.

Equation (16a) needs to be complemented with corresponding initial and boundary conditions. In the sequel of this section, we focus the discussion on the case of a two dimensional macroscopic domain, i.e.  $x = (x_1, x_2) \in [0, 1] \times [0, 1]$ .

We set Robin conditions at the one side of the square

$$\frac{\partial u_i}{\partial n}(t, x_1, 0) + b_r u_i(t, x_1, 0) = \begin{cases} u_i^b(x_1) > 0 & t \in [0, t_0], \\ 0 & t > t_0, \end{cases}, \quad x_1 \in [0, 1], \tag{16f}$$

while we impose Neumann boundary conditions for the rest of the boundary

$$\frac{\partial u_i}{\partial n}(t, x_1, x_2) = 0, (16g)$$

for  $(x_1, x_2)$  such that  $0 \le x_2 \le 1$  with  $x_1 = 0, 1$  or  $0 \le x_1 \le 1$  with  $x_2 = 0$  and with initial conditions

$$u_i(x,0) = u_{i0}(x) \ge 0.$$
 (16h)

Moreover, we have for  $\beta = \sum_{i=1}^{N} b_i$ ,

$$\partial_t v(t,x) = \sum_{i=1}^N \alpha_i u_i(t,x) - \beta v(t,x), \tag{16i}$$

with some initial condition

$$v(x,0) = v_0(x) \ge 0, (16j)$$

and

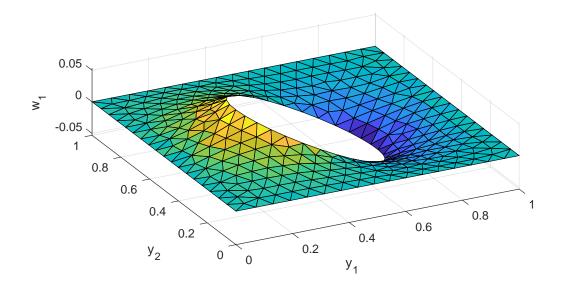
$$r(t,x)\,\partial_t r(t,x) = \alpha \left(\sum_{i=1}^N a_i u_i(t,x) - \beta v(t,x)\right) L(t,x),\tag{16k}$$

together with some initial distribution

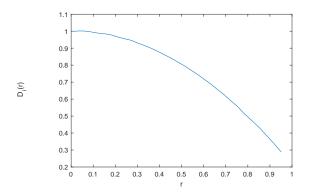
$$r(x,0) = r_0(x) > 0, (161)$$

for  $x\in[0,1]\times[0,1]$ . We discuss in Section 4.2 additional choices of suitable initial and boundary conditions. Note also that the restriction  $r_0(x)>0$  is posed strictly for modelling reasons. Taking  $r_0(x)=0$  does not create any particular problems in the previous theoretical analysis or in the numerical treatment of the problem that follows. Although since the model refers to a porous medium which consists of an initial solid skeleton we impose  $r_0(x)>0$ , to model such a consideration. In summary the solution strategy is to solve initially the cell problem, (16d), (16e) for a sufficient number of values of r in  $[r_0,1]$ . In such a way and via interpolation and numerical differentiation for the evaluation of  $\nabla w_k$  we can obtain approximations of  $D_i(r)$  for every  $r\in[0,1]$ . Due to the geometry of the domain in the cell problem a finite element method is used. Next we apply a forward in time, finite difference scheme in order to solve the macroscopic equations (16a), (16i) for the concentrations  $u_i$  and v, while at each time step r is given by a discretization of (16k) allowing next for the evaluation of the terms  $L,A,D_i$  at each point (t,x) of the macroscopic domain. Since we consider a square macroscopic domain the finite difference scheme, instead of a more advanced as a finite element one, is used and in order to gain some simplicity in the implementation of it.

Such a solution strategy for the same problem, but for a one dimensional macroscopic domain is also used in [26], and for a similar system of equations in [9].



**Figure 3:** Numerical solution of the cell problem (16d) and specifically for  $w_1$  with  $r_i = .25$ .



**Figure 4:** The diffusion coefficient dependence (for  $d_i = 1$ ) on the radius r.

### 4.2 Discretization schemes

To treat problem (16) numerically, we need to obtain firstly a numerical approximation for the cell problems (16d) and determine the shape of the corresponding cell functions  $w_1, w_2$  posed in  $Y_0(t,x)$ .

More specifically, we proceed for the various values of r, for  $r_0 \le r(t,x) \le 1/2$ . We take a partition of width  $\delta r$ ,  $r_0, r_1 = r_0 + \delta r, \dots, r_{M_1} = 1/2$ .

Then since  $Y_0$  is determined as the area contained inside the square cell and outside the circle of radius r, we obtain a sequence of solutions for the cell problem (16d) for each  $Y_{0i}$  corresponding to the radius  $r_i$  of the partition.

We use a finite element scheme to solve these cell problems. To be precise, we use the MATLAB finite element package "Distmesh" (see details in [31]) to triangulate the domain  $Y_{0i} = Y_0(r_i)$ . Furthermore, a solver has been implemented to handle this specific problem (equations (16d)); it works in a similar fashion as applied in [26].

In Figure 3, we illustrate the numerical solution for this problem for a particular choice of  $r_i$ . Specifically, we choose to look at  $r_i=.25$ . Moreover in Figure 4 it is plotted the diffusion coefficient  $D_i(r)$  (for  $d_i=1$ ) against the radius r. The form of The dependence of the diffusion coefficient is in accordance with similar results as for example in [25], [9], [42]) Having available the numerical evaluation of the cell functions  $w_k$  as approximate solutions to the cell problems (16d) and (16e),

the entries of the diffusion tensor  $(D_i)_{jk}=d_i\int_{Y_0(t,x)}\left(\delta_{j,k}+\nabla_{y_j}w_k\right),\ i=1,\ldots,N,\ j,k=1,2$  can be calculated directly and for each (t,x) and consequently for the corresponding value for r(t,x) and thus for  $Y_0(t,x)$ . Then the corresponding value of  $(D_i)_{jk}(t,x)$  is approximated via linear interpolation.

Next, we solve the system of equations (16a)-(16l). We use a finite difference scheme to solve the two-dimensional version of the field equation (16a), together with its boundary and initial conditions. More specifically we consider a square domain  $\Omega = [0,1] \times [0,1]$ .

For this purpose we implement a forward finite difference scheme and for this purpose initially we consider a uniform partition of the domain  $\Omega$ , with  $x=(x_1,\,x_2)\in\Omega$ ,  $0\leq x_1\leq 1$ ,  $0\leq x_2\leq 1$ , of  $(M+1)\times(M+1)$  points with spacial step  $\delta x_1=\delta x_2=\delta x$ , with  $x_{1\ell_1}=\ell_1\delta x$ ,  $\ell_1=0,1,\ldots M$ ,  $x_{2\ell_2}=\ell_2\delta x$ ,  $\ell_2=0,1,\ldots M$ .

Additionally, we take a partition of  $N_T$  points in the time interval [0,T], where T is the maximum time of the simulation, with step  $\delta t$  and  $t_n=n\delta t,\,i=0,\ldots N_T-1$ .

Let  $U_{i\ell_1,\ell_2}^n$  the numerical approximation of the species i of the solution of equation (16a) at the point  $(t_n,x_{1\ell_1},x_{2\ell_2})$  of  $\Omega_T=\Omega\times[0,T]$ , that is  $u_i(t_n,x_{1\ell_1},x_{2\ell_2})\simeq U_{i\ell_1,\ell_2}^n$ . Moreover we denote by  $\mathrm{D}_{i\ell_1,\ell_2}^n$  the corresponding approximation of the diffusion coefficients  $(D_i)_{jk}(t_n,x_{1\ell_1},x_{2\ell_2})\simeq \mathrm{D}_{i\ell_1,\ell_2}^n$  and similarly by  $V_{i\ell_1,\ell_2}^n$  the approximation for the species  $v,v(t_n,x_{1\ell_1},x_{2\ell_2})\simeq V_{\ell_1,\ell_2}^n$ .

Finite difference scheme for the model equations. Initially we focus on the appropriate discretization of the terms in (16a). For the spatial derivatives  $\frac{\partial}{\partial x_s}\left(D_i(t,x)\frac{\partial u_i}{\partial x_s}\right)$ , where s=1,2 we apply a discretization of the form

$$\begin{split} \frac{\partial}{\partial x_1} \left( D_i(t,x) \frac{\partial u_i}{\partial x_1} \right) &\simeq \Delta \left( u_i(D_i u_{ix_1}) \right)_{x_1} := \frac{1}{\delta x} \left[ \mathbf{D}_{i\ell_1 + \frac{1}{2},\ell_2}^n \left( \frac{U_{i\ell_1 + 1,\ell_2}^n - U_{i\ell_1,\ell_2}^n}{\delta x} \right) - \mathbf{D}_{i\ell_1 - \frac{1}{2},\ell_2}^n \left( \frac{U_{i\ell_1,\ell_2}^n - U_{i\ell_1 - 1,\ell_2}^n}{\delta x} \right) \right] \\ \frac{\partial}{\partial x_2} \left( D_i(t,x) \frac{\partial}{\partial x_2} \right) &\simeq \Delta \left( u_i(D_i u_{ix_2}) \right)_{x_2} := \frac{1}{\delta x} \left[ \mathbf{D}_{i\ell_1,\ell_2 + \frac{1}{2}}^n \left( \frac{U_{i\ell_1,\ell_2 + 1}^n - U_{i\ell_1,\ell_2}^n}{\delta x} \right) - \mathbf{D}_{i\ell_1,\ell_2 - \frac{1}{2}}^n \left( \frac{U_{i\ell_1,\ell_2 - U_{i\ell_1,\ell_2 - 1}}^n}{\delta x} \right) \right] \\ & \mathbf{D}_{i\ell_1 + \frac{1}{2},\ell_2} = \frac{\mathbf{D}_{i\ell_1 + 1,\ell_2} + \mathbf{D}_{i\ell_1,\ell_2}}{2}, \quad \mathbf{D}_{i\ell_1 - \frac{1}{2},\ell_2} = \frac{\mathbf{D}_{i\ell_1,\ell_2} + \mathbf{D}_{i\ell_1,\ell_2 - 1}}{2}, \\ & \mathbf{D}_{i\ell_1,\ell_2 + \frac{1}{2}} = \frac{\mathbf{D}_{i\ell_1,\ell_2 + 1} + \mathbf{D}_{i\ell_1,\ell_2}}{2}, \quad \mathbf{D}_{i\ell_1,\ell_2 - \frac{1}{2}} = \frac{\mathbf{D}_{i\ell_1,\ell_2} + \mathbf{D}_{i\ell_1,\ell_2 - 1}}{2}. \end{split}$$

Moreover we use a standard forward in time discretization for the time derivative and we conclude with a finite difference scheme of the form for the species  $u_i$ 's,

$$U_{i\ell_{1},\ell_{2}}^{n+1} = U_{i\ell_{1},\ell_{2}}^{n} + \delta t \, \Delta \left( U_{i}(Du_{ix_{1}}) \right)_{x_{1}} + \delta t \, \Delta \left( U_{i}(Du_{ix_{1}}) \right)_{x_{1}} + \delta t R_{i\ell_{1},\ell_{2}}^{n} - \delta t F_{\ell_{1},\ell_{2}}^{n}$$

and for the species v

$$V_{\ell_1,\ell_2}^{n+1} = V_{\ell_1,\ell_2}^n + \delta t \sum_{i=1}^N \alpha_i U_{i\ell_1,\ell_2}^n - \beta V_{\ell_1,\ell_2}^n,$$

where

$$R_{i\ell_{1},\ell_{2}}^{n} = \frac{1}{2} \sum_{p+q=s} \gamma_{p,q} U_{p\ell_{1},\ell_{2}}^{n} U_{q\ell_{1},\ell_{2}}^{n} - U_{s\ell_{1},\ell_{2}}^{n} \sum_{p=1}^{N-s} \gamma_{s,p} U_{p\ell_{1},\ell_{2}}^{n},$$

and

$$F_{\ell_1,\ell_2}^n = \frac{L_{\ell_1,\ell_2}^n}{A_{\ell_1,\ell_2}^n} \left( a_i U_{i\ell_1,\ell_2}^n - b_i V_{\ell_1,\ell_2}^n \right),$$

are the approximations of the source terms at the point  $(t_n, x_{1\ell_1}, x_{2\ell_2})$ .

In addition, the functions for the length L(r) and for the area A(r), are approximated, for  $r \le 1/2$  by the relations:

$$L^n_{\ell_1,\ell_2} = 2\pi r^n_{\ell_1,\ell_2}, \quad A^n_{\ell_1,\ell_2} = 1 - \pi (r^n_{\ell_1,\ell_2})^2, \quad \text{(in 2D)}.$$

Furthermore, we have the approximate value  $r^n_{\ell_1,\ell_2}$  of the radius r given by

$$r_{\ell_1,\ell_2}^{n+1} = r_{\ell_1,\ell_2}^n + \delta t \frac{1}{r_{\ell_1,\ell_2}^n} \alpha \left( \sum_{i=1}^N a_i U_{i\ell_1,\ell_2}^n - \beta V_{\ell_1,\ell_2}^n \right) L_{\ell_1,\ell_2}^n.$$

The scheme has been tested regarding its convergence and the expected second order convergence in the  $L^2$  norm is attainable. Moreover the stability of the scheme is also supported experimentally by ensuring that in each experiment  $D_i(r)\frac{\delta_t}{\delta^2} \leq 0.25$ .

## 4.3 Basic simulation output

In the first set of simulations we consider homogeneous Neumann boundary conditions at the three edges of the square  $\Omega$ , namely at  $x_1=0, x_1=1$  for  $0 \le x_2 \le 1$  and at  $x_2=1, 0 \le x_1 \le 1$ .

At the edge  $x_2=0,\,0\leq x_1\leq 1$  we impose Robin boundary conditions given by equation (16f). That is we consider a scenario of having inflow at this side of  $\Omega$  for a particular time period,  $[0,t_0]$  which stops after some time  $t_0$ , and we want mainly to observe the deposition process of the colloid species around the solid cores of the cells. The later is apparent by the variation in time of the radius r.

We take zero distributions as initial conditions (t=0) for the colloidal populations, while we consider various specific initial distributions for the radius r.

We consider N=3 mobile species  $u_i$  and one immobile species v. Our model needs a quite large number of parameters. These dimensionless parameters were chosen in a range indicated by the model evaluation in [20], [22, 26] . We take them as follows:  $(d_1, d_2, d_3) = (.3, .5, .99)$ ,  $(a_1, a_2, a_3) = (.9, .5, .3)$ ,  $(\beta_1, \beta_2, \beta_3) = (1, 1, 1)$ ,  $\alpha_{i,j} = .1$ ,  $\beta_{i,j} = 100$ ,  $i, j = 1, \ldots 3$ ,  $u_{i0}(x) = 0$ .

Regarding the choice of boundary condition at  $(x_1,0)$ , we take the function  $u_i^b$  to be defined as

$$(u_1^b, u_2^b, u_3^b) = (u_{10}^b x_1(1-x_1), 0, 0)$$

with  $u_{10}^b=25$  for  $t\in[0,t_0]$  and zero for  $t>t_0$ , with  $t_0=2$ . Moreover, we let  $b_r=0.5$ ,  $v(0,x_1,x_2)=0$ , and a uniform initial distribution of the radius r,  $r(0,x_1,x_2)=r_0(x_1,x_2)=0.1$ .

In addition, we take as final simulation time T=3 and set the remaining parameters to be  $M=41,\,{\rm R}:=\delta t/\delta x^2=0.25.$ 

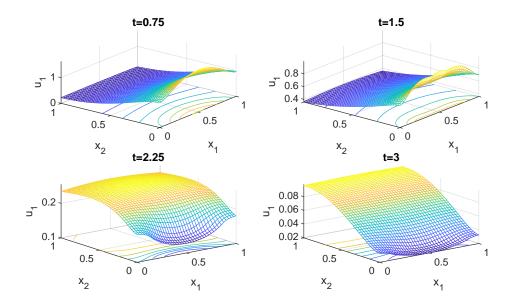
**Approximated concentration profiles.** In the first of the following graphs, i.e. in Figure 5, concentration profiles of the colloidal population  $u_1$  are plotted against space. Similar profiles are exhibited by the other colloidal populations as well. As general rule, we keep the discussion about what happens with  $u_1$  only as here the effects are more visible. This corresponds also to the physical situation when most of the mass is contained in the monomer population, while the amount of observable dimer, trimer, 4-mer populations is considerably lower; see e.g. [22] and references cited therein.

In the first two frames we have  $t < t_0$ ; hence we can see that there is an inflow in  $\Omega$  through one edge and so we can observe the diffusion of  $u_1$  taking place in the  $x_2$  direction. In the last two frames taken at times after  $t_0$  (hence here the inflow has stopped) we see that the concentration of  $u_1$  near the edge drops possibly due to an activation of the reaction mechanisms. Especially, the deposition activates and consumes monomers initially involved in diffusion.

In Figure 6, we present similar graph for the concentration of  $u_2$ . As expected, the behaviour is similar as for the species  $u_1$ . Moreover, for the third species  $u_3$  during the simulation we notice no difference in its qualitative behaviour.

Regarding the behaviour of the immobile species v pointed out in Figure 7, we observe an initial distribution in the first two frames  $t=0.75,\,t=1.5$ , following the form of the mobile species  $u_i$  and an increase inside the domain  $\Omega$ . After the inflow stops, for instance, see the last two frames  $t=2.25,\,t=3$ , the distribution of the mass of the deposited species appears to be stationary.

Focusing now in the behaviour of r, we present in Figure 8 time frames of contour plots of the radius at times  $t_i = 0.75, 1.5, 2.25, 3$ . We observe the expected increase of the radius with respect



**Figure 5:** Concentration profiles at different time steps for the species  $u_1$ .

to time. Even for  $t>t_0=2$ , after the inflow has stopped to happen, we still have a slight increase of the radius due to the accumulation of the immobile species around the spherical cores of the cells. Moreover close at the end of the simulation, T=3 we do not observe any significant changes in the system. This is also indicated by Figure 9.

As final remarks regarding this numerical experiment, the main observables  $u_1$ ,  $u_2$ ,  $u_3$ , and v are plotted in Figure 9 against time for fixed locations inside the domain  $\Omega$ ; see specifically the points (0,0.5), the center (0.5,0.5), (0.5,1) and at the corner (0,0). These points are chosen indicatively at the boundary and at the center of the domain so that to observe the behaviour of the species concentrations at these points. The discontinuity from the inflow at  $x_1=0$  apparent at the point (0,0) smooths out in the inner points of the domain while the maximum of the concentrations are depicted almost at the same time instant.

**Approximations with non uniform initial radius.** In the following experiment we consider for the same scenario of initial and boundary conditions, (16f), (16g), (16h), a non uniform distribution for the initial values of the radius  $r_0 = r(0, x_1, x_2)$ . Specifically, we consider larger values of the radius in the form of two peaks centered at the points (0.2, 0.2) and (0.8, 0.8) and with  $r_0$  having the form

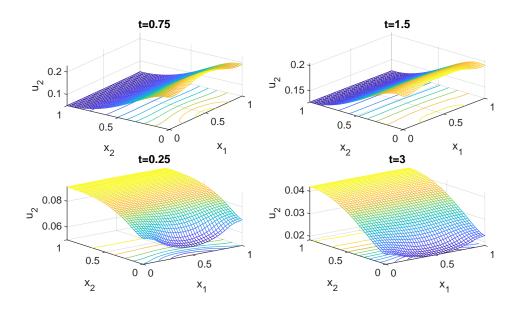
$$r_0 = r_c + r_1 \exp\left[-c(x_1 - .2)^2 - c(x_2 - .2)^2\right] + r_1 \exp\left[-c(x_1 - .8)^2 - c(x_2 - .8)^2\right].$$

This is the initial distribution also demonstrated in Figure 13 (left) of section 2.

In this context, we take  $r_c=0.05,\,r_1=0.35,\,c=60$  so that the maximum radius at these two points is quite large but smaller than one  $(\max r(0,x_1,x_2)\simeq 0.42)$  as it can be seen in the yellow area shown in Figure 11. Here we also set M=41 for the spatial partition and R=0.25, The rest of the parameters values are the same as in the previous numerical experiment.

The effect of the non-uniform initial radius distribution is apparent in the evolution of the species of the model; particularly, this non-uniformity effect can be traced back in the evolution of the population  $u_1$  as exhibited in Figure 10.

Due to the inflow from the edge  $x_2 = 0$ , we have now high values in the  $u_1$  concentration around this edge (yellow area) of the domain, while inside the domain we have lower value (blue areas);



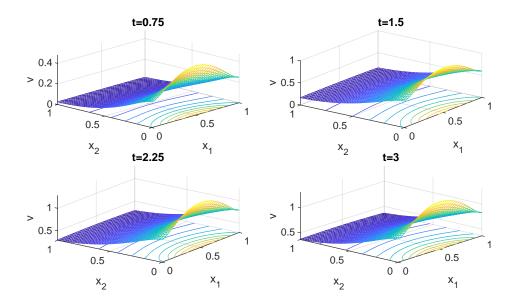
**Figure 6:** Concentration profiles at different time steps for the species  $u_2$ .

this behavior can be seen in the first two frames of the simulation  $(t=0.75,\,t=1.5)$ . We notice a gradual increasing perturbation of the symmetric form of  $u_1$  around the point  $(0.2,\,0.2)$  due to the fact that, precisely at this point, we have large values of r. In the next frames, at  $(t=2.25,\,t=3)$  and particularly at t=2.25, we observe the concentration of  $u_1$  after the time that the inflow in the domain has stopped  $(t>t_0$  and  $\frac{\partial u_i}{\partial n}(t,x_1,0)+b_ru_i(t,x_1,0)=0)$ . The dominant mechanisms now are the diffusion and the surface reaction, i.e. the deposition of material around the cores of the cells. Thus we observe lower values of  $u_1$  (blue and green areas) around the points with larger r (close to the two initial peaks of r) where there the material has been deposited and higher values (yellow areas) in between the aforementioned peak points where the values of r are smaller and deposition is slower. Essentially due to the same mechanism, at the final frame t=3 at the end of the simulation, the values of  $u_1$  decrease and tend to zero with slower speed within the area close to the corner (0,1).

In Figure 11, we present the contour plot of the initial value of r for this experiment. In Figure 12, we point out the spatial distribution of the radius  $r=r(T,x_1,x_2)$ , where T is the final time of the simulation. In this case, we observe a behaviour consistent with what happens with the profile of the colloidal population  $u_1$  towards the end of the simulation, i.e. around t=3. This effect is shown in Figure 10.

Higher values of r equal to 0.5, where clogging occurs, are taken in the lower part of the domain near the edge  $x_2=0$  as well as in the neighbour of the points  $(0.2,\,0.2)$  and  $(0.8,\,0.8)$ ; observe the yellow areas in Figure 12. In the rest of the domain  $\Omega$  the radius r attains lower values. This is in line with the observed behaviour of the concentration profiles of  $u_1$  around the end of the simulation we observe higher values of the concentration  $u_1$  around the edge (0,1) where we have the area with the lower values of r.

The evolution of the diffusivity during the experiment is also apparent in Figure 13. We notice initially low values of it in the areas (blue regions) around the two peaks and higher values in the intermediate area (yellow region), in the first frame for t=0.75. As r gradually increases the corresponding areas with low diffusivity expand as we can see in the second and third frame for  $t=1.5,\,2.25$ , and finally, also for t=3 at the end of the simulation where we obtain the final map of



**Figure 7:** Mass at different time steps for the deposited species v.

the diffusivity. The latter frame is in fact a "reverse" image of Figure 12 as very low values of D are linked to clogging around the blue areas where r is large.

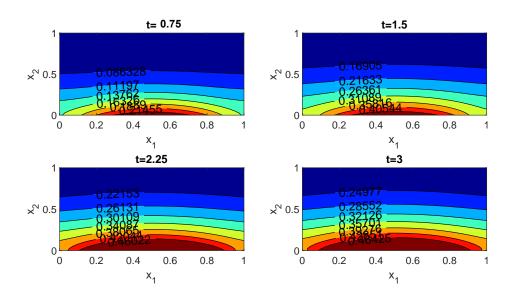
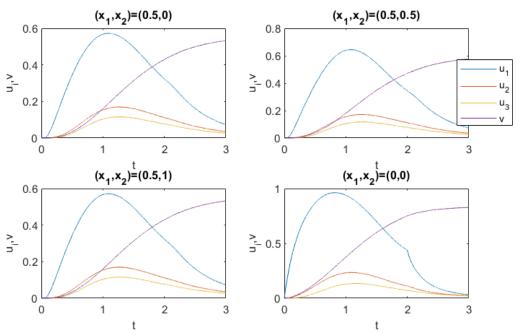
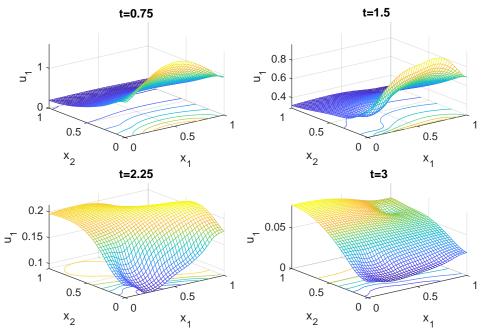


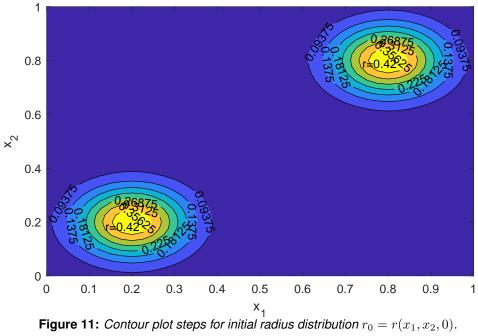
Figure 8: Contour plots of the radius  $r=r(x_1,x_2,t_i)$  for the time steps  $t_i=0.75,\,1.5,\,2.25,\,3.$ 

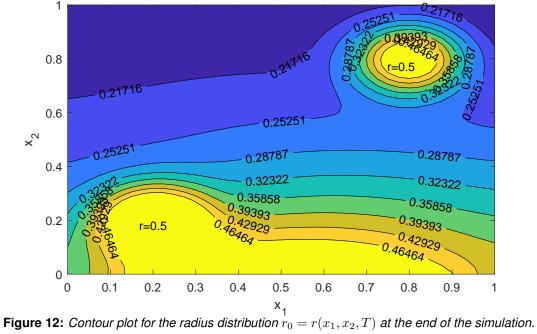


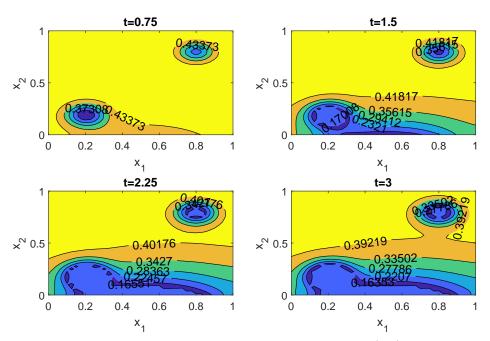
**Figure 9:** Concentration profiles of the species  $u_i$ , v versus time at different spatial points in the square domain.



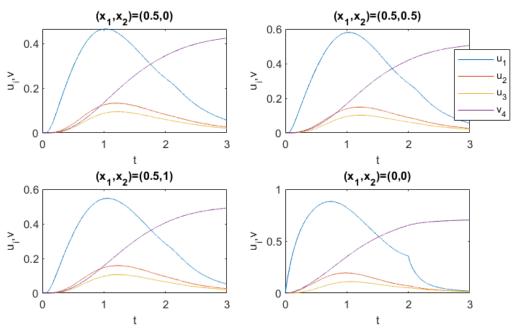
**Figure 10:** Mesh plots at different time steps for the concentration of the species  $u_1$  for the case of nonuniform initial radius distribution.







**Figure 13:** Contour plots at different time steps for the effective diffusivity D(t,x) for the case of nonuniform initial radius distribution.



**Figure 14:** Concentration profiles of the species  $u_i$ , v versus time at different spatial points in the square domain in the case of non uniform initial radius distribution.

It is worthwhile to note that the spatial distribution of the balls-like microstructure that corresponds to the vizualization shown in Figure 12 of the effective transport coefficient is pointed out in Figure 1. The unavoidable occurrence of clogging is pointed out in all these representations. Note that since we consider a two dimensional setting in our model and at this context by clogging we define the situation when neighbouring cycles touch each other i.e. when r=0.5. Locally we still have non solid areas between the intersecting circles and thus we have still positive diffusion.

Finally, in agreement with Figure 9, we present the concentrations  $u_1$ ,  $u_2$ ,  $u_3$ , and v in Figure 14 against time for fixed locations picked inside the domain  $\Omega$  at the the points (0.5,0), the center (0.5,0.5), (0.5,1) and at the corner (0,0). The observed behaviour is similar as in the experiment with uniform radius distribution. Note tough that we have a steeper decrease of profile of the species  $u_1$  at these points and a lower maximum at the point (0.5,0). This is a direct consequence of the non-uniformity of the initial radius distribution.

### 5 Discussion

We have proven the existence of a weak solution to a specific coupled multiscale quasilinear system describing the diffusion, aggregation, and deposition of populations of colloidal particles in porous media. The structure of the system was originally derived in [26] and we kept it here.

Tracking numerically the x-dependence in the shape of the microstructures rises serious computational problems, this would especially be the case in a 3D scenario or when working with shapes of lower regularity instead of circles.

Because of the strong separation between the macroscopic length scale and the microscopic length scale, such settings are parallelizable; see [37] for a prestudy in this direction done for a micro-macro reaction-diffusion problem with x-dependent microstructure arising in the context of transport of nutrients in plants. The approach used in [37] is potentially applicable in our scenario as well and might be worthwhile in more complex scenarios where the general shape of the porous medium (circles in our case) is not prescribed but itself part of the overall problem.

Moreover, what concerns the discretization techniques used in this framework, a more advanced finite difference scheme, such as an appropriate version of Du Fort Frankel scheme, can give in principle more flexibility and accuracy in the numerical computations, e.g. by allowing larger time steps.

Our multiscale model can allow for further relevant extensions in at least twofold direction:

- (1) For instance, a particularly interesting development would be to allow for some amount of stochasticity in the balance laws. In this spirit, the ODE for the growth of the balls induced by the deposition of the species v could have not only a random distribution of initial positions<sup>3</sup> but also some suitably scaled "Brownian noise" in the production term mimicking an additional contribution eventually due to a non-uniform deposition of colloids on the boundary of the microstructures (compare with the setting from [3]). The difficulty in this case is that, due to the strong coupling in the system, the overall problem becomes a quasilinear SPDE, which is much more difficult to handle mathematically and from the simulation point of view compared with our current purely deterministic setting.
- (2) Another development that would be interesting to follow in the deterministic setup is to attempt a computational efficient hybrid-type modeling. In this context, one idea would be to couple continuum population models for colloidal dynamics with discrete network models describing the mechanics of the underlying material (see e.g. the approach proposed in [21] having paper as target material). Relevant questions would be: What is the counterpart of our equation for the radius growth of a ball B(r), when the continuously-distributed ball is replaced by a finite number of lower dimensional objects like points and interfaces modeling defects and fibers, respectively? How does "continuum" deposition take place on such "discrete" fixed locations?

We expect that the non-standard type of couplings suggested in (1) and (2) (i.e. deterministic-stochastic and continuum-discrete) can potentially be posed in terms of measured-valued balance

<sup>&</sup>lt;sup>3</sup>This is tractable with the current form of the model.

equations. We will investigate some of these ideas in follow-up works.

# **Acknowledgments**

The authors are grateful to the anonymous referees for their constructive input that helped to considerable improve the initial manuscript.

AM is partially supported by the grant VR 2018-03648 "Homogenization and dimension reduction of thin heterogeneous layers". We thank R. E. Showalter (Oregon) and O. Richardson (Karlstad) for useful discussions on closely related topics.

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