

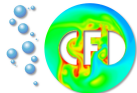
# Efficient, accurate and flexible Finite Element solvers for Chemotaxis

BIOMATH 2011, 15-18 June 2011, Sofia

Robert Strehl   Andriy Sokolov   Stefan Turek

Technische Universität Dortmund  
Institut für Angewandte Mathematik/Numerik, LS III

June 16, 2011



- 1 underlying models
- 2 numerical challenges
- 3 final notes

- 1 underlying models
- 2 numerical challenges
- 3 final notes

**Chemotaxis** describes an oriented movement towards or away from regions of higher concentrations of chemical agents and plays a vitally important role in the evolution of many living organisms.

<http://dictybase.org/Multimedia/motility/motility.htm>

It is common to use continuous models  $\rightarrow$  system of partial differential equations (PDE)

Minimal Keller-Segel model (1970) for chemotaxis:

$$\begin{array}{ll}
 \text{equation for motile} & \frac{\partial u}{\partial t} = \nabla \cdot \left( \underbrace{\nabla u}_{\text{diffusion}} - \underbrace{\chi u \nabla c}_{\text{chemotaxis}} \right) \\
 \text{species } u: & \\
 \text{equation for the} & \frac{\partial c}{\partial t} = \underbrace{\Delta c}_{\text{diffusion}} - \underbrace{c + u}_{\text{reaction}} \\
 \text{chemical agent } c: &
 \end{array}$$

Since 1970 various models have been proposed (especially in the recent decades).

$$\frac{\partial u}{\partial t} = \nabla \cdot \left( \underbrace{\nabla u}_{\text{diffusion}} - \underbrace{\chi u \nabla c}_{\text{chemotaxis}} \right)$$

$$\frac{\partial c}{\partial t} = \underbrace{\Delta c}_{\text{diffusion}} - \underbrace{c + u}_{\text{reaction}}$$

Since 1970 various models have been proposed (especially in the recent decades).

$$\begin{aligned}
 \frac{\partial u}{\partial t} &= \nabla \cdot \left( \underbrace{D(u)\nabla u}_{\text{diffusion}} - \underbrace{\chi(u, c)\nabla c}_{\text{chemotaxis}} \right) \\
 \frac{\partial c}{\partial t} &= \underbrace{\Delta c}_{\text{diffusion}} - \underbrace{\alpha c + \beta(u)u}_{\text{reaction}}
 \end{aligned}$$

(nonlinear) coefficients modeling e.g.  $D(u), \chi(u, c), \beta(u) \xrightarrow{u \rightarrow \infty} 0$   
saturation effects:

Since 1970 various models have been proposed (especially in the recent decades).

$$\begin{aligned}
 \frac{\partial u}{\partial t} &= \nabla \cdot \left( \underbrace{D(u)\nabla u}_{\text{diffusion}} - \underbrace{\chi(u, c)\nabla c}_{\text{chemotaxis}} \right) + \underbrace{f(u)}_{\text{kinetics}} \\
 \frac{\partial c}{\partial t} &= \underbrace{\Delta c}_{\text{diffusion}} - \underbrace{\alpha c + \beta(u)u}_{\text{reaction}}
 \end{aligned}$$

(nonlinear) coefficients modeling e.g.  $D(u), \chi(u, c), \beta(u) \xrightarrow{u \rightarrow \infty} 0$   
saturation effects:

introducing kinetics: e.g.  $f(u) = \nu u(1 - u)$  (logistic)



Since 1970 various models have been proposed (especially in the recent decades).

$$\frac{\partial u_i}{\partial t} = \nabla \cdot \left[ \left( \sum_{l=1}^N D_{i,l}^u(u_i) \nabla u_l \right) - \left( \sum_{k=1}^M \chi_{i,k}(u_i) \nabla c_k \right) \right] + f_i(u_i)$$

$$\frac{\partial c_j}{\partial t} = D_j^c \Delta c_j - \sum_{k=1}^M \alpha_{k,j} c_k + \sum_{l=1}^N \beta_{l,j} u_l$$

(nonlinear) coefficients modeling saturation effects: e.g.  $D(u), \chi(u, c), \beta(u) \xrightarrow{u \rightarrow \infty} 0$

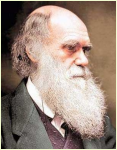
introducing kinetics:

e.g.  $f(u) = \nu u(1 - u)$  (logistic)

multispecies:

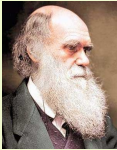
e.g. species  $u_1, \dots, u_N$ , chemical agents  $c_1, \dots, c_M$

## Biology



- models are well motivated
- all ingredients for their own are well understood

## Biology



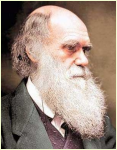
- models are well motivated
- all ingredients for their own are well understood

## Mathematics

- existence and uniqueness are nontrivial
- analysis revealed mathematical artifacts



## Biology



- models are well motivated
- all ingredients for their own are well understood

→ numerical ansatz is highly desired to validate models and obtain more insights from mathematical point of view

## Mathematics

- existence and uniqueness are nontrivial
- analysis revealed mathematical artifacts



1) the minimal model may lead to blowing up solutions. From biological point of view, those unbounded solutions do not make any sense.

## minimal model

$$\frac{\partial u}{\partial t} = \nabla \cdot (\nabla u - \chi u \nabla c)$$

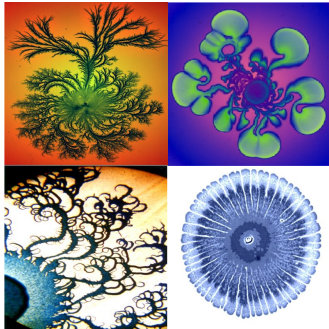
$$\frac{\partial c}{\partial t} = \Delta c - c + u$$

$\mathbb{R}^1$  : all solutions are bounded

$\mathbb{R}^2$  : blow-up iff  $\|u_0\|_1 > 8\pi/\chi$

$\mathbb{R}^{\geq 3}$  : no explicit threshold is known

2) Stunning results were obtained when biologists study certain mutated bacteria colonies. Their proliferation seems to follow certain patterns.



E. Ben-Jacob,  
<http://star.tau.ac.il/~eshel/image-flow.html>

## kinetic model

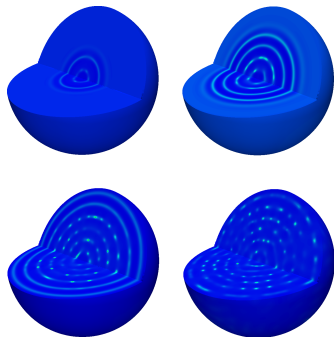
$$\begin{aligned}\frac{\partial u}{\partial t} &= \nabla \cdot (Du - \chi u \nabla c) + \nu u(1 - u) \\ \frac{\partial c}{\partial t} &= \Delta c - \beta c + u\end{aligned}$$

$\mathbb{R}^{1,2}$  : unique global weak solution  
(at least for  $\nu \gg 1$ )

$\mathbb{R}^{\geq 3}$  : far less is known

existence of nontrivial steady states

2) Stunning results were obtained when biologists study certain mutated bacteria colonies. Their proliferation seems to follow certain patterns.



## kinetic model

$$\begin{aligned}\frac{\partial u}{\partial t} &= \nabla \cdot (Du - \chi u \nabla c) + \nu u(1 - u) \\ \frac{\partial c}{\partial t} &= \Delta c - \beta c + u\end{aligned}$$

$\mathbb{R}^{1,2}$  : unique global weak solution  
(at least for  $\nu \gg 1$ )

$\mathbb{R}^{\geq 3}$  : far less is known  
existence of nontrivial steady states

- 1 underlying models
- 2 numerical challenges
- 3 final notes



In order to obtain a reliable solver for chemotaxis PDEs many (numerical) concerns has to be tackled:

### challenges

- high-order resolution (of sharp interfaces/steep gradients)
- fast solver techniques
- smart memory management
- robustness for a variety of parameters
- user interface (arbitrary coefficients)
- mass conservation (when applicable) and positivity preservation

In order to obtain a reliable solver for chemotaxis PDEs many (numerical) concerns has to be tackled:

### challenges

- high-order resolution (of sharp interfaces/steep gradients)
- fast solver techniques
- smart memory management
- robustness for a variety of parameters
- user interface (arbitrary coefficients)
- mass conservation (when applicable) and positivity preservation

Especially the last three are of particular interest in the presence of chemotaxis PDEs.

Applying standard (high-order) Finite Element Methods (FEM) on chemotaxis dominated PDEs lead to severe numerical instabilities. When restricted to the minimal model, the troublemaker is the essential chemotaxis term  $\nabla \cdot (\chi u \nabla c)$ .

Applying standard (high-order) Finite Element Methods (FEM) on chemotaxis dominated PDEs lead to severe numerical instabilities. When restricted to the minimal model, the troublemaker is the essential chemotaxis term  $\nabla \cdot (\chi u \nabla c)$ .

→ upwind schemes guarantee to 'smooth-out' instabilities and preserve physical entities

Applying standard (high-order) Finite Element Methods (FEM) on chemotaxis dominated PDEs lead to severe numerical instabilities. When restricted to the minimal model, the troublemaker is the essential chemotaxis term  $\nabla \cdot (\chi u \nabla c)$ .

→ upwind schemes guarantee to 'smooth-out' instabilities and preserve physical entities



**BUT: high-order is not anymore obtained.**

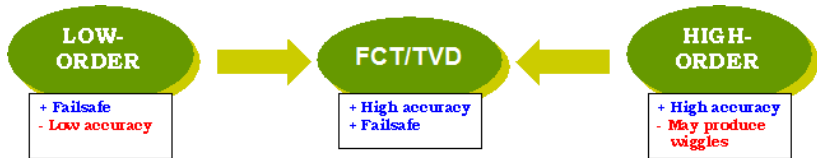
Applying standard (high-order) Finite Element Methods (FEM) on chemotaxis dominated PDEs lead to severe numerical instabilities. When restricted to the minimal model, the troublemaker is the essential chemotaxis term  $\nabla \cdot (\chi u \nabla c)$ .

→ upwind schemes guarantee to 'smooth-out' instabilities and preserve physical entities



**BUT:** high-order is not anymore obtained.

**REMEDY:** merging the two approaches leads to FCT/TVD which combines all desired properties



In the presence of more comprehensive models, the introduced nonlinearities also ask for a special treatment.

Common segregated linearization techniques converge very poorly when applied to ill-conditioned systemmatrices.

a segregated approach:

$$1. A_1(u_{n-1}, c_{n-1}) c_n = b_1$$

$$2. A_2(u_{n-1}, c_n) u_n = b_2$$

In the presence of more comprehensive models, the introduced nonlinearities also ask for a special treatment.

Common segregated linearization techniques converge very poorly when applied to ill-conditioned systemmatrices.

set up a block system matrix (monolithic approach) and apply (damped) Newton-like or fixpoint methods

a segregated approach:

$$1. A_1(u_{n-1}, c_{n-1}) c_n = b_1$$

$$2. A_2(u_{n-1}, c_n) u_n = b_2$$

a monolithic approach:

$$\underbrace{\begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}}_{=: A(u_n, c_n)} \begin{pmatrix} u_n \\ c_n \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$$



When developing a software for solving a diversity of underlying models, an user-prescribed input is highly favorable.

Our software FEAST/FEATFLOW is designed in a module based fashion and allows for easy access via single 'stand-alone' objects.

When developing a software for solving a diversity of underlying models, an user-prescribed input is highly favorable.

Our software FEAST/FEATFLOW is designed in a module based fashion and allows for easy access via single 'stand-alone' objects.

## Generic super-model

The current underlying generic (single-species) model reads:

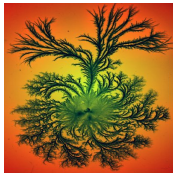
$$\begin{aligned}\frac{\partial u}{\partial t} &= \nabla \cdot (D(u) \nabla u - \chi(u, c) \nabla c) + f(u) \\ \frac{\partial c}{\partial t} &= \Delta c - \alpha c + \beta(u) u\end{aligned}$$

→ all coefficients may be user-prescribed

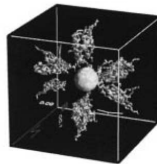
- 1 underlying models
- 2 numerical challenges
- 3 final notes

Certainly, applied mathematicians look for practical benefits of their work. Since chemotaxis plays a key role for organisms, plenty applications come into mind.

- proliferation of bacteria (not only in petri dishes)
- tumour growth/angiogenesis/haptotaxis
- breeding concerns (insemination of sea urchins)
- immunology (production of chemokines at infection sites)



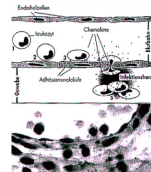
E. Ben-Jacob,  
[http://star.tau.ac.il/~eshel/  
image-flow.html](http://star.tau.ac.il/~eshel/image-flow.html)



M.A.J. Chaplain,  
Journal of Neuro-Oncology



C. Pietschmann, MPI



L. Kinzel, LMU

The developed software embeds the following features:

- supported domains:  $\Omega \subset \mathbb{R}^2, \mathbb{R}^3$  (reasonable mesh restrictions)
- spatial discretization via  $Q_1, Q_2, \dots$  elements
- temporal discretization:  $\theta$ -scheme
- reasonable boundary conditions at will: Dirichlet, Neumann, periodic,...
- user-prescribed parameters/coefficients/callback functions (module-based Open Source Software)
- FCT/TVD stabilized solver (preservation of physical entities)
- embedded nonlinear solvers: (Deuflhard) damped Newton-like methods, fixpoint, Picard-linearization
- graphical output via GMV/PARAVIEW

Further aims for the software:

- extend the framework to multi-species systems
- implementation of fast multigrid-solvers
- spatial (h-, r-) and temporal (t-) adaptivity
- parallelization
- ...

## Further informations:

- email: `robert.strehl@math.tu-dortmund.de`
- homepage: `http://www.mathematik.tu-dortmund.de/~rstrehl/downloads.html`
- software: `http://www.feathflow.de`
- model organism: `http://dictybase.org`

**list of figures:** `http://dictybase.org/Multimedia/motility/motility.htm` ; `http://www.youtube.com/watch?v=hpHpBHJZQvU` ;

`http://star.tau.ac.il/~eshel/image-flow.html` ; M. A. J. Chaplain, Mathematical modelling of angiogenesis, Journal of Neuro-Oncology, Vol.

50, pp. 37-51, 2000 ; Catarina Pietschmann, MaxPlanckForschung 2009 Heft 2, Wo, bitte, geht's denn hier zum Ei? ; Linda Kinzel,

Seminar Autoimmunität, Einführung Chemokine, 24./25. Juni 2006