

Computational Studies of Bacterial Colony Model

Ondřej Pártl

Department of Mathematics, Faculty of Nuclear Sciences and Physical Engineering, Czech Technical University in Prague, Prague, Czech Republic Email: partlond@fjfi.cvut.cz

Received January 17, 2013; revised February 24, 2013; accepted April 1, 2013

Copyright © 2013 Ondřej Pártl. This is an open access article distributed under the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

ABSTRACT

Microbiological experiments show that the colonies of the bacterium bacillus subtilis placed on a dish filled with an agar medium and nutrient form varied spatial patterns while the individual cells grow, reproduce and migrate on the dish in clumps. In this paper, we discuss a system of reaction-diffusion equations that can be used with a view to modelling this phenomenon and we solve it numerically by means of the method of lines. For the spatial discretization, we use the finite difference method and Galerkin finite element method. We present how the spatial patterns obtained depend on the spatial discretization employed and we measure the experimental order of convergence of the numerical schemes used. Further, we present the numerical results obtained by solving the model in a cubic domain.

Keywords: Bacterial Colony Model; Reaction-Diffusion Equations; Method of Lines; Galerkin Finite Element Method

1. Introduction

According to microbiological experiments, the colonies of the bacterium *bacillus subtilis* placed on a dish filled with an agar medium and nutrient form varied—and surprisingly regular—patterns while the individual cells grow, reproduce and migrate on the dish in clumps (see, e.g., [1,2]). As bacterium is an extremely primitive organism, the question arises whether the motion of the cells is governed by some simple underlying principles that can be described by mathematically.

The general approach to the pattern formation in biology is based on continuous and discrete dynamical models or neural models as described in [3]. In this case, the simple character of the motion of the bacteria resembles the Brownian motion and indicates the analogy with the diffusion limited aggregation (see [4]). Detailed analysis of the results of the experiments coupled the diffusion effects with the reactive behaviour of the species and led to the derivation of the model based on the reaction-diffusion system (see [5,6]) which was studied in [2,7]. In this article, we discuss it with respect to the accuracy and reliability of the numerical solution.

The model is based on the assumption that the cells can be divided into two classes: the active and the inactive. The active cells, with concentration u(x,t) at the point x and time t, grow, move, feed on the nutrient and reproduce; whereas the inactive ones, with concentration

w(x,t), do not do anything. Finally, let v(x,t) be the concentration of the nutrient.

Then the time and space evolution of the concentrations within a bounded spatial domain $\Omega \subset \mathbb{R}^n$, with n=2 or n=3, is governed by the system

$$\frac{\partial u}{\partial t} = \mathrm{d}\Delta u + uv - a(u, v)u,\tag{1}$$

$$\frac{\partial v}{\partial t} = \Delta v - uv,\tag{2}$$

$$\frac{\partial w}{\partial t} = a(u, v)u \tag{3}$$

for $x \in \Omega$ and t > 0, where d represents a diffusion coefficient, and the function a = a(u, v) is of the form

$$a(u,v) = \frac{a_0}{\left(1 + \frac{u}{a_1}\right)\left(1 + \frac{v}{a_2}\right)},$$

with positive constants a_0, a_1, a_2 . The system (1)-(3) is subject to the initial conditions

$$u(x,0) = u_0(x),$$

$$v(x,0) = v_0(x),$$

$$w(x,0) \equiv 0$$
(4)

for $x \in \Omega$ where $0 \le u_0, v_0 \le M$ are $C(\overline{\Omega})$ functions $(\overline{\Omega})$ represents the closure of Ω), and M is a positive

constant; and the boundary conditions

$$\frac{\partial u}{\partial n} = 0, \frac{\partial v}{\partial n} = 0, \quad x \in \partial \Omega, t > 0,$$

where $\partial\Omega$ denotes the boundary of Ω , and $\frac{\partial u}{\partial n}$ stands for the derivative of u in the direction of the unit outward normal.

This problem can be written in the following weak formulation:

$$\frac{\mathrm{d}}{\mathrm{d}t}(u,\psi) = -\mathrm{d}(\nabla u, \nabla \psi) + (uv - a(u,v)u,\psi), \quad (5)$$

$$\frac{\mathrm{d}}{\mathrm{d}t}(v,\tilde{\psi}) = -(\nabla v, \nabla \tilde{\psi}) - (uv,\tilde{\psi}) \tag{6}$$

for all $\psi, \tilde{\psi}$ from the Sobolev space $W_2^1(\Omega)$ in the sense of distributions on a time interval I=(0,T), T>0, where (\cdot,\cdot) stands for the inner product

$$(\psi, \tilde{\psi}) = \int_{\Omega} \psi(x) \tilde{\psi}(x) dx$$
, where $\psi, \tilde{\psi} \in W_2^1(\Omega)$;

$$\frac{\mathrm{d}}{\mathrm{d}t}w = a(u,v)u\tag{7}$$

for all $x \in \overline{\Omega}$ in the sense of distributions on I; and

$$u(0) = u_0, \ v(0) = v_0, \ w(0) = w_0,$$
 (8)

where $u_0, v_0, w_0 \in L_2(\overline{\Omega})$.

The existence and uniqueness of the weak solution are proved in [8]. The authors also proved that

$$\lim_{t\to+\infty} \left(u(t),v(t)\right) = \left(0,v_{\infty}\right), \text{ uniformly in } \Omega,$$

for a positive constant v_{∞} , and that there is a function w_{∞} (a final colony pattern) such that

$$\lim_{t \to +\infty} w(x,t) = w_{\infty}(x), \text{ uniformly in } \overline{\Omega}.$$

Figure 1 shows the observed types of the spatial pat-

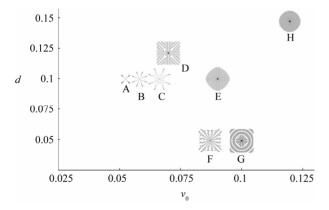


Figure 1. Types of the spatial patterns in $\Omega \subset \square^2$ produced by the model studied. The patterns D-H correspond to the results of microbiological experiments. The results A-C have not been found in literature.

terns in $\Omega \subset \mathbb{R}^2$ (the darker points correspond to the higher concentration of the cells) that have been found by the author by solving system (1)-(3).

2. Numerical Schemes

The model under consideration was solved numerically by means of the method of lines; for the spatial discretization, the finite difference method and Galerkin finite element method (the latter only in the case $\Omega \subset \mathbb{R}^2$) were used. For the sake of simplicity, we use only square and cubic domains Ω , specifically

$$\Omega = (-L, L) \times (-L, L)$$

and

$$\Omega = (-L, L) \times (-L, L) \times (-L, L), L > 0.$$

2.1. Finite Difference Method

In the two-dimensional case, the domain Ω is covered by a regular orthogonal mesh of $(N+1)^2$ nodes (x_i, y_j) ,

with the step
$$h = \frac{1}{N}, N \in \square$$
. Let us denote

 $u(x_i, y_j, t) = u_{i,j}(t)$ ($v_{i,j}(t)$ and $w_{i,j}(t)$ are defined similarly). The spatial derivatives in (1) and (2) are approximated by the second order central difference, and the derivatives in the boundary conditions are approximated by the first order backward difference. Therefore, we obtain the following system of ODE's:

$$\frac{du_{i,j}(t)}{dt} = \frac{d}{h^2} \left[u_{i+1,j}(t) + u_{i-1,j}(t) - 4u_{i,j}(t) + u_{i,j+1}(t) + u_{i,j-1}(t) \right]
+ u_{i,j}(t) v_{i,j}(t) - a \left(u_{i,j}(t), v_{i,j}(t) \right) u_{i,j}(t),
\frac{dv_{i,j}(t)}{dt}
= \frac{1}{h^2} \left[v_{i+1,j}(t) + v_{i-1,j}(t) - 4v_{i,j}(t) + v_{i,j+1}(t) + v_{i,j-1}(t) \right]
- u_{i,j}(t) v_{i,j}(t),
\frac{dw_{i,j}(t)}{dt} = a \left(u_{i,j}(t), v_{i,j}(t) \right) u_{i,j}(t),$$

where in the first two equations, $i, j = 1, 2, 3, \dots, N-1$; and in the third equation, $i, j = 0, 1, 2, \dots, N$. Further, on the boundary we get

$$\begin{split} u_{0,j}\left(t\right) &= u_{1,j}\left(t\right), & v_{0,j}\left(t\right) &= v_{1,j}\left(t\right); \\ u_{N,j}\left(t\right) &= u_{N-1,j}\left(t\right), & v_{N,j}\left(t\right) &= v_{N-1,j}\left(t\right); \\ u_{j,0}\left(t\right) &= u_{j,1}\left(t\right), & v_{j,0}\left(t\right) &= v_{j,1}\left(t\right); \\ u_{j,N}\left(t\right) &= u_{j,N-1}\left(t\right), & v_{j,N}\left(t\right) &= v_{j,N-1}\left(t\right); \end{split}$$

for $j = 1, 2, 3, \dots, N-1$;

$$\begin{aligned} u_{0,0}\left(t\right) &= u_{1,1}\left(t\right), & u_{0,N}\left(t\right) &= u_{1,N-1}\left(t\right), \\ u_{N,0}\left(t\right) &= u_{N-1,1}\left(t\right), & u_{N,N}\left(t\right) &= u_{N-1,N-1}\left(t\right); \\ v_{0,0}\left(t\right) &= v_{1,1}\left(t\right), & v_{0,N}\left(t\right) &= v_{1,N-1}\left(t\right), \\ v_{N,0}\left(t\right) &= v_{N-1,1}\left(t\right), & v_{N,N}\left(t\right) &= v_{N-1,N-1}\left(t\right). \end{aligned}$$

This system is supplemented with the initial conditions

$$u_{i,j}(0) = u_0(x_i, y_j), v_{i,j}(0) = v_0(x_i, y_j),$$

 $w_{i,j}(0) = w_0(x_i, y_j),$

where $i, j = 0, 1, 2, \dots, N$.

In the case of the cubic domain $\,\Omega$, the spatial discretization is derived analogically. These two schemes will be called MOL.

2.2. Galerkin Finite Element Method

In this method, both the linear (triangular) and the bilinear (square) Lagrange elements (see [9]) are used. The basis $\left\{\xi_{j}\right\}_{j=1}^{N_{h}}$ of the finite dimensional subspace $S_{h} \subset$

 $W_2^1(\Omega)$ consists of the functions which are linear on each triangle (in the first case) or bilinear on each square (in the second case) and take the value 1 at one node of the spatial mesh and vanish at the other nodes. In the first case, two types of triangulations are used; they are depicted in **Figure 2**. In the case of the square finite elements, the same spatial mesh as in the method of lines is utilized. Further, the integrand in $(a(u,v)u,\psi)$ in (5) is approximated by its interpolant (see [9]). Finally, the numerical scheme is simplified employing the method of lumped masses (see [10,11]).

Hence, writing

$$u(t) = \sum_{j=1}^{N_h} u_j(t) \xi_j \text{ and } v(t) = \sum_{j=1}^{N_h} v_j(t) \xi_j,$$
 (9)

where N_h is the number of the nodes of the spatial mesh, Equations (5)-(7) are transformed as follows:

$$\frac{\mathrm{d}}{\mathrm{d}t}u_{k}(t)\int_{\Omega}\xi_{k}^{2}\mathrm{d}x = -\mathrm{d}\sum_{j=1}^{N_{h}}u_{j}(t)\int_{\Omega}(\nabla\xi_{k})(\nabla\xi_{j})\mathrm{d}x
+ \int_{\Omega}\left(\sum_{i=1}^{N_{h}}u_{i}(t)\xi_{i}\right)\left(\sum_{j=1}^{N_{h}}v_{j}(t)\xi_{j}\right)\xi_{k}\mathrm{d}x
- \int_{\Omega}I_{h}\left\{a\left(\sum_{j=1}^{N_{h}}u_{j}(t)\xi_{j},\sum_{j=1}^{N_{h}}v_{j}(t)\xi_{j}\right)\left(\sum_{i=1}^{N_{h}}u_{i}(t)\xi_{i}\right)\xi_{k}\right\}\mathrm{d}x,$$
(10)

$$\frac{\mathrm{d}}{\mathrm{d}t}v_{k}(t)\int_{\Omega}\xi_{k}^{2}\mathrm{d}x = -\sum_{j=1}^{N_{h}}v_{j}(t)\int_{\Omega}(\nabla\xi_{k})(\nabla\xi_{j})\mathrm{d}x$$

$$-\int_{\Omega}\left(\sum_{i=1}^{N_{h}}u_{i}(t)\xi_{i}\right)\left(\sum_{j=1}^{N_{h}}v_{j}(t)\xi_{j}\right)\xi_{k}\mathrm{d}x,$$
(11)

$$\frac{\mathrm{d}}{\mathrm{d}t} w_k(t) = a(u_k(t), v_k(t)) u_k(t) \tag{12}$$

for $k = 1, 2, 3, \dots, N_h$, where I_h stands for the interpolant.

This system is supplemented with the initial conditions

$$u_k(0) = u_0(x_k, y_k, 0), \quad v_k(0) = v_0(x_k, y_k, 0),$$

 $w_k(0) = w_0(x_k, y_k, 0),$

where $k = 1, 2, 3, \dots, N_h$.

The schemes corresponding to the cases depicted in **Figure 2** will be denoted FEMHS, FEMRHS and FEMSS, respectively.

The convergence of the numerical schemes derived can be proved as in [12-14]. The systems of ODE's derived were solved by means of the Runge-Kutta-Merson method (see [12,15]) with the adaptive time step control.

3. Quantitative Studies

In the two-dimensional case, the experimental order of convergence (EOC), see [16], was measured; only the error of the functions u and v was considered. As the model studied does not possess an analytical solution, Equations (1) and (2) were transformed into the form

$$\frac{\partial u}{\partial t} = d\Delta u + uv - a(u, v)u + f,$$

$$\frac{\partial v}{\partial t} = \Delta v - uv + g,$$
(13)

where the additional terms f = f(x, y, t) and g = g(x, y, t) assure that the system possesses the solution

$$u(x, y, t) = \sin(t)\sin(kx)\sin(ly) + \gamma,$$

$$v(x, y, t) = \sin(t)\sin(kx)\sin(ly) + \delta,$$

where k, l, γ, δ are positive constants.

Therefore, it turns out that

$$f(x, y, t)$$

$$= \sin(kx)\sin(ly)\left[\cos(t) + d\sin(t)(k^2 + l^2)\right]$$

$$-u(x, y, t)v(x, y, t) + a(u(x, y, t), v(x, y, t))u(x, y, t),$$

$$g(x, y, t) = \sin(kx)\sin(ly)\left[\cos(t) + \sin(t)(k^2 + l^2)\right]$$

$$+u(x, y, t)v(x, y, t)$$

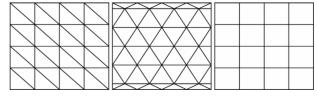


Figure 2. Meshes used for the Galerkin FEM.

and $k = l = \frac{\pi}{2L}$ (from the boundary conditions).

The modified system was solved analogically as the original one (the functions f and g were approximated in the same way as the functions u and v in (9)). The EOC coefficient α is defined by

$$\frac{\left\|z(t) - z_{h_1}(t)\right\|}{\left\|z(t) - z_{h_2}(t)\right\|} = \left(\frac{h_1}{h_2}\right)^{\alpha},\tag{14}$$

where $z = (u, v)^T$, z_{h_i} represents the numerical solution corresponding to the parameter h_i which denotes the constant distance between two nodes on the bottom side (according to **Figure 2**) of the spatial mesh used (there are N nodes on this side), and $\|\cdot\|$ denotes a suitable norm. The following form of Equation (14) was used:

$$\frac{\max_{\substack{\{vi\}_{j=1}^{T} \\ \{vi\}_{j=1}^{T} }} \left\{ \left\| z(vi) - z_{h_{1}}(vi) \right\| \right\}}{\max_{\substack{\{vi\}_{j=1}^{T} \\ \{vi\}_{j=1}^{T} }} \left\{ \left\| z(vi) - z_{h_{2}}(vi) \right\| \right\}} = \left(\frac{h_{1}}{h_{2}}\right)^{\alpha}, \tag{15}$$

where T stands for a final time, and ν is a time step between the compared time levels of the solution. Two norms $||z(t)-z_h(t)||$ were used:

$$\begin{split} & \left\| z(t) - z_{h}(t) \right\|_{L_{2}(\Omega)} \\ &= \left(\left\| u(t) - u_{h}(t) \right\|_{L_{2}(\Omega)}^{2} + \left\| v(t) - v_{h}(t) \right\|_{L_{2}(\Omega)}^{2} \right)^{\frac{1}{2}}, \\ & \left\| z(t) - z_{h}(t) \right\|_{L_{\infty}(\Omega)} \\ &= \left\| u(t) - u_{h}(t) \right\|_{L_{\infty}(\Omega)} + \left\| v(t) - v_{h}(t) \right\|_{L_{\infty}(\Omega)}, \end{split}$$

with

$$\left\|u(t)-u_h(t)\right\|_{L_2(\Omega)} = \sqrt{\sum_i h_{\text{ref}}^2 \left|u^i(t)-u_h^i(t)\right|^2},$$

$$\left\|u(t)-u_h(t)\right\|_{L_{\infty}(\Omega)} = \max_i \left\{\left|u^i(t)-u_h^i(t)\right|\right\},$$

where $u^i(t), u^i_h(t)$ denote the nodal values of the projections of u(t) and $u_h(t)$ onto a regular orthogonal mesh with a spatial step h_{ref} .

The parameters were chosen as follows: L=1, $a_0=1$, $a_1=\frac{1}{2400}$, $a_2=\frac{1}{120}$, d=0.1, $\gamma=1$ and $\delta=1$. Further, T=10 (for the colored FEMPHS, FEMPHS)

ther, T = 10 (for the schemes FEMRHS, FEMHS and FEMSS) or T = 5 (for the scheme MOL), $\nu = 1$; and the mesh for the evaluation of the error had 4001×4001 nodes ($h_{\rm ref} = 0.0005$).

The results are summarized in **Tables 1-4**. The EOC coefficients in the i-th row are computed from the data in the i-th and (i-1)-th row. The schemes FEMRHS, FEMHS and FEMSS exhibit the second order of convergence in h, and MOL exhibits the first order. Further,

Table 1. Results of the EOC measurement for the scheme MOL.

MOL					
N	h	error $L_{\scriptscriptstyle \!$	error L_2	EOC $L_{\scriptscriptstyle \!$	EOC L ₂
51	0.040000	0.529063	0.359874	-	-
101	0.020000	0.272997	0.178844	0.954555	1.008790
151	0.013333	0.184121	0.119209	0.971336	1.000361
201	0.010000	0.138648	0.089272	0.986087	1.005317
251	0.008000	0.111235	0.071390	0.987218	1.001756
301	0.006667	0.093114	0.059582	0.975576	0.991976
351	0.005714	0.079951	0.051065	0.988096	0.999921
401	0.005000	0.069768	0.044574	1.020631	1.018478

Table 2. Results of the EOC measurement for the scheme FEMSS.

		F	EMSS		
N	h	error $L_{_{\!$	error L_2	EOC $L_{\scriptscriptstyle \!$	EOC L ₂
51	0.040000	0.077088	0.050120	-	-
101	0.020000	0.018893	0.012299	2.028651	2.026828
151	0.013333	0.008375	0.005448	2.006352	2.008133
201	0.010000	0.004706	0.003061	2.003423	2.004120
251	0.008000	0.003002	0.001958	2.015349	2.002681
301	0.006667	0.002084	0.001359	2.002393	2.002250
351	0.005714	0.001613	0.000999	1.659161	1.998986
401	0.005000	0.001173	0.000764	2.385993	2.002004

Table 3. Results of the EOC measurement for the scheme FEMHS.

		F	EMHS		
N	h	error $L_{\scriptscriptstyle \!$	error L_2	EOC $L_{\scriptscriptstyle \!$	EOC L ₂
51	0.040000	0.158644	0.105160	-	-
101	0.020000	0.037661	0.025031	2.074658	2.070788
151	0.013333	0.016615	0.011027	2.018022	2.021731
201	0.010000	0.009307	0.006184	2.014809	2.010393
251	0.008000	0.005990	0.003953	1.974826	2.006225
301	0.006667	0.004136	0.002743	2.031748	2.005075
351	0.005714	0.003031	0.002014	2.014271	2.001729
401	0.005000	0.002330	0.001542	1.972873	2.002282

Table 4. Results of the EOC measurement for the scheme FEMRHS.

		FE	MRHS		
N	h	error $L_{\scriptscriptstyle \!$	error L ₂	EOC $L_{\scriptscriptstyle \!$	EOC L ₂
51	0.040000	0.083246	0.054366	-	-
101	0.020000	0.019875	0.013025	2.066438	2.061396
151	0.013333	0.008724	0.005726	2.030527	2.026983
201	0.010000	0.004880	0.003206	2.019736	2.015895
251	0.008000	0.003113	0.002047	2.014464	2.010246
301	0.006667	0.002157	0.001420	2.011809	2.007288
351	0.005714	0.001583	0.001043	2.007380	2.002991
401	0.005000	0.001226	0.000803	1.911572	1.957965

FEMSS appears to be the most accurate of all.

4. Computational Studies of Pattern Formation

This section provides the comparison of final spatial patterns obtained by the four schemes under consideration in a square domain. All the results were computed for L = 500, $a_0 = 1$, $a_1 = \frac{1}{2400}$ and $a_2 = \frac{1}{120}$ (similarly as

in the articles [2] or [7]); and the initial conditions were of the form

$$u_0(x, y) = 0.4 \cdot \exp\{-10^{-5} \cdot (x^4 + y^4)\},\$$

 $w_0(x, y) \equiv 0.$

The function v_0 was chosen constant. Therefore, the pattern types vary depending on the values of the parameters v_0 and d.

The results presented (**Figures 3-11**) have the form of coloured pictures with a final shape of the function w, computed by one of the schemes, in shades of grey; each of the coloured lines in these pictures traces the spatial shape of the function w computed by one of the schemes for one value of N. Further, in **Tables 5-7** there are differences between these shapes expressed by means of the L_{∞} - and L_2 -norm (plus one another table for the combination $v_0 = 0.25$, d = 0.25). The value of the L_2 -norm is divided by the area of Ω . The time evolution of the patterns is shown in [17].

Note that, in general, it cannot be stated that one scheme is more or less accurate than the other schemes because, typically, each of them produces another shape. In author's opinion, the most significant difference between the schemes consists in the symmetry of the results obtained by the scheme FEMHS.

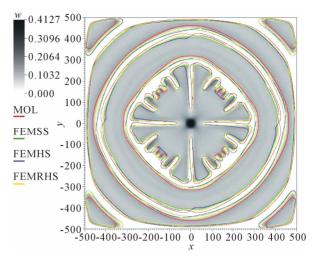


Figure 3. Comparison of the schemes for $v_0 = 0.1$, d = 0.05 and N = 1601.

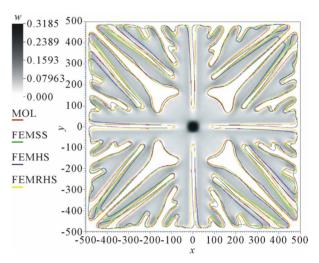


Figure 4. Comparison of the schemes for $v_0 = 0.071$, d = 0.12 and N = 1601.

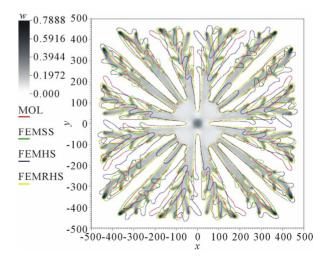


Figure 5. Comparison of the schemes for $v_0 = 0.087$, d = 0.05 and N = 1601.

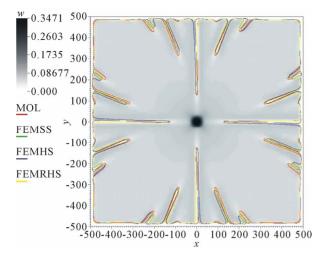


Figure 6. Comparison of the schemes for $v_0 = 0.09$, d = 0.1 and N = 1601.

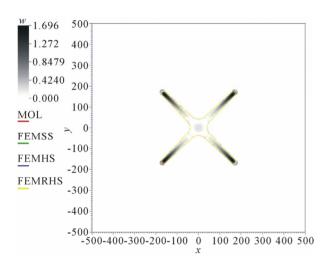


Figure 7. Comparison of the schemes for $v_0 = 0.055$, d = 0.1 and N = 1601.

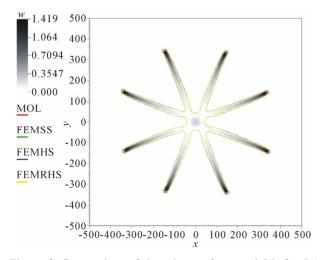


Figure 8. Comparison of the schemes for $v_0 = 0.06$, d = 0.1 and N = 1601.

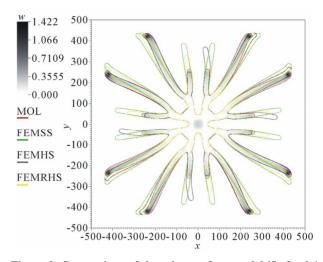


Figure 9. Comparison of the schemes for $v_0 = 0.065$, d = 0.1 and N = 1601.

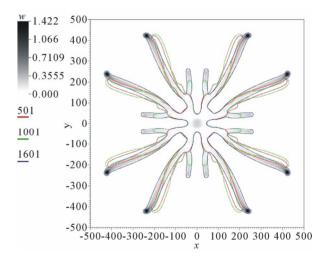


Figure 10. Comparison of the patterns computed by the scheme MOL for $v_0 = 0.065$, d = 0.1 and different values of N.

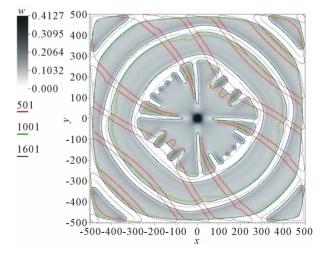


Figure 11. Comparison of the patterns computed by the scheme FEMHS for $v_0 = 0.1$, d = 0.05 and different values of N.

Table 5. Comparison of the final shapes of the function w. The shapes compared are depicted in Figures 3-6. In the case $v_0 = 0.25$, d = 0.25, the results are too similar to be compared graphically.

v_0	= 0.1, d = 0.05, N = 16	501
schemes	difference $L_{\scriptscriptstyle \!$	difference L_2
MOL:FEMSS	1.9667×10 ⁻¹	4.6688×10 ⁻⁵
MOL:FEMHS	2.4377×10^{-1}	4.3610×10 ⁻⁵
MOL:FEMRHS	1.9978×10^{-1}	3.2744×10 ⁻⁵
FEMSS:FEMHS	1.9668×10^{-1}	2.5633×10 ⁻⁵
FEMSS:FEMRHS	1.3298×10^{-1}	2.0264×10^{-5}
FEMHS:FEMRHS	2.0281×10^{-1}	2.5920×10 ⁻⁵
$v_0 =$	0.071, d = 0.12, N = 1	601
schemes	difference $L_{\scriptscriptstyle \!$	difference L_2
MOL:FEMSS	1.2000×10 ⁻¹	2.5271×10 ⁻⁵
MOL:FEMHS	1.1659×10^{-1}	3.0695×10 ⁻⁵
MOL:FEMRHS	1.0848×10^{-1}	1.4082×10 ⁻⁵
FEMSS:FEMHS	1.3434×10^{-1}	2.8857×10 ⁻⁵
FEMSS:FEMRHS	1.2010×10^{-1}	1.6973×10 ⁻⁵
FEMHS:FEMRHS	1.2617×10^{-1}	2.9721×10 ⁻⁵
$v_0 =$	0.087, d = 0.05, N = 1	.601
schemes	difference $L_{\scriptscriptstyle \!$	difference L_2
MOL:FEMSS	8.0538×10^{-1}	1.0412×10 ⁻⁴
MOL:FEMHS	7.5287×10^{-1}	1.2401×10^{-4}
MOL:FEMRHS	7.8884×10^{-1}	1.1858×10 ⁻⁴
FEMSS:FEMHS	8.0536×10^{-1}	1.2202×10 ⁻⁴
FEMSS:FEMRHS	8.0510×10^{-1}	1.0690×10 ⁻⁴
FEMHS:FEMRHS	7.8884×10^{-1}	1.0980×10^{-4}
v_0	= 0.09, d = 0.1, N = 16	501
schemes	difference $L_{\scriptscriptstyle \!$	difference L_2
MOL:FEMSS	6.2300×10^{-2}	5.3668×10 ⁻⁶
MOL:FEMHS	8.4376×10^{-2}	7.6022×10^{-6}
MOL:FEMRHS	5.0979×10^{-2}	4.4019×10^{-6}
FEMSS:FEMHS	7.7855×10^{-2}	7.2132×10^{-6}
FEMSS:FEMRHS	1.9048×10^{-2}	1.3308×10^{-6}
FEMHS:FEMRHS	7.5321×10^{-2}	7.2604×10^{-6}
$v_0 =$	= 0.25, d = 0.25, N = 1	601
schemes	difference $L_{\scriptscriptstyle \!$	difference L
MOL:FEMSS	3.1984×10^{-3}	2.8760×10 ⁻⁷
MOL:FEMHS	3.1058×10^{-3}	2.7736×10^{-7}
MOL:FEMRHS	2.9924×10^{-3}	2.2338×10 ⁻⁷
FEMSS:FEMHS	1.0946×10^{-3}	1.0970×10 ⁻⁷
FEMSS:FEMRHS	6.5345×10^{-4}	7.6301×10^{-8}
FEMHS:FEMRHS	7.8323×10^{-4}	1.0060×10^{-7}

Table 6. Comparison of the final shapes of the function w. The shapes compared are depicted in Figures 7-9.

$v_0 = 0.55, \ d = 0.1, \ N = 1601$					
schemes	difference $L_{_{\!$	difference L_2			
MOL:FEMSS	3.4125×10^{-1}	9.0358×10^{-6}			
MOL:FEMHS	6.4508×10^{-1}	1.3066×10^{-5}			
MOL:FEMRHS	2.9847×10^{-1}	7.8921×10^{-6}			
FEMSS:FEMHS	3.1521×10^{-1}	8.2631×10^{-6}			
FEMSS:FEMRHS	4.3125×10^{-2}	1.3477×10^{-6}			
FEMHS:FEMRHS	3.5638×10^{-1}	8.5530×10^{-6}			
v ₀ =	$v_0 = 0.06, \ d = 0.1, \ N = 1601$				
schemes	difference $L_{_{\!$	difference L_2			
MOL:FEMSS	1.4660	6.0531×10 ⁻⁵			
MOL:FEMHS	1.4187	5.8731×10 ⁻⁵			
MOL:FEMRHS	1.4181	5.8869×10^{-5}			
FEMSS:FEMHS	$4.4940\!\times\!10^{^{-1}}$	1.7464×10^{-5}			
FEMSS:FEMRHS	$5.0207\!\times\!10^{-2}$	$2.0685\!\times\!10^{-6}$			
FEMHS:FEMRHS	$4.2681\!\times\!10^{^{-1}}$	1.7111×10 ⁻⁵			
$v_0 =$	0.065, d = 0.1, N = 1	601			
schemes	difference $L_{\scriptscriptstyle\!$	difference L_2			
MOL:FEMSS	1.4219	1.1143×10^{-4}			
MOL:FEMHS	1.3156	7.0282×10^{-5}			
MOL:FEMRHS	1.3946	9.4370×10^{-5}			
FEMSS:FEMHS	1.1538	1.0058×10^{-4}			
FEMSS:FEMRHS	9.5829×10^{-1}	7.4929×10^{-5}			
FEMHS:FEMRHS	1.0295	7.0227×10^{-5}			

Table 7. Comparison of the final shapes of the function w computed by one scheme with different values of N. The shapes compared are depicted in Figures 10 and 11.

	$v_0 = 0.065, d = 0.1$		
N	difference $L_{\scriptscriptstyle \!$	difference L_2	
501:1001	8.1949×10^{-1}	3.4029×10^{-5}	
501:1601	9.2009×10^{-1}	3.8776×10^{-5}	
1001:1601	1.9787×10^{-1}	$8.1497\!\times\!10^{^{-6}}$	
	$v_0 = 0.1, d = 0.05$		
N	difference $L_{\scriptscriptstyle \!$	difference L_2	
501:1001	1.9079×10^{-1}	5.7197×10^{-5}	
501:1601	$1.8888\!\times\!10^{^{-1}}$	6.5279×10^{-5}	
1001:1601	1.7536×10^{-1}	4.8792×10^{-5}	

5. Three-Dimensional Pattern Formation

This section provides examples of the spatial patterns that arise if we consider the cubic domain with L = 200 or L = 300. Similar types of the initial conditions and values of the parameters as in the two-dimensional case

were used, *i.e.*,
$$a_0 = 1$$
, $a_1 = \frac{1}{2400}$ and $a_2 = \frac{1}{120}$;
 $u_0(x, y, z) = 0.4 \cdot \exp\{-10^{-5} \cdot (x^4 + y^4 + z^4)\}$,
 $w_0(x, y, z) \equiv 0$,
 $v_0(x, y, z) \equiv \text{const.}$

(z denotes the third spatial coordinate). The values of L, N, v_0 and d represented variable parameters of the model.

The results are presented in **Figures 12-25**. They have the form of coloured isolines of the functions u, v or u+w. The time evolution is shown only for one combination of parameters; in the other cases, the solution evolves analogically.

6. Conclusions

It follows from the numerical results that the shape of the numerical solution strongly depends on the spatial discretization employed. The pattern type, nevertheless, seems to be stable, if the spatial mesh is not too coarse. In author's opinion, therefore, the numerical methods under consideration can be used for the investigation of the pattern growth described by the system of the reaction-diffusion equations discussed.

As for the measurement of the experimental order of convergence, it is remarkable that the method of lumped masses does not seem to decrease it. Finally, the three-dimensional model exhibits very complex behaviour as well.

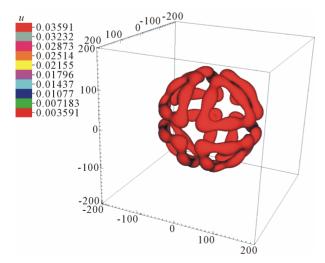


Figure 12. Isolines of u for L = 200, N = 201, $v_0 = 0.06$ and d = 0.1 at t = 2500.

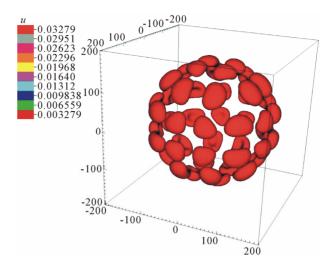


Figure 13. Isolines of u for L = 200, N = 201, $v_0 = 0.06$ and d = 0.1 at t = 3500.

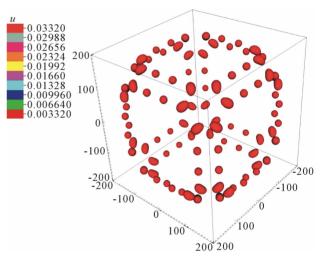


Figure 14. Isolines of u for L = 200, N = 201, $v_0 = 0.06$ and d = 0.1 at t = 6500.

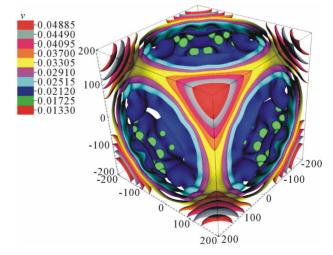


Figure 15. Isolines of v for L = 200, N = 201, $v_0 = 0.06$ and d = 0.1 at t = 6500.

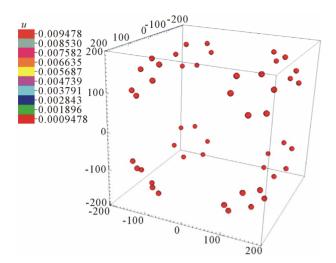


Figure 16. Isolines of u for L = 200, N = 201, $v_0 = 0.06$ and d = 0.1 at t = 8000.

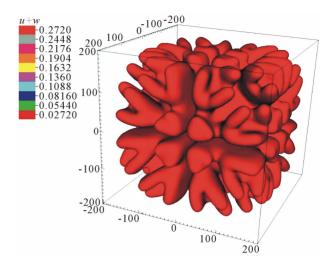


Figure 19. Isolines of u + w for L = 200, N = 201, $v_0 = 0.065$, d = 0.1 and t = 9000.

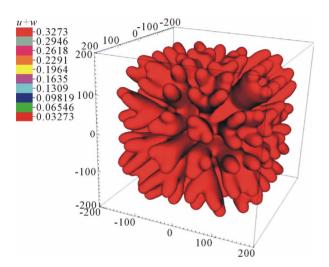


Figure 17. Isolines of u+w for $L=200,\,N=201,\,\nu_0=0.06$ and d=0.1 at t=8000.

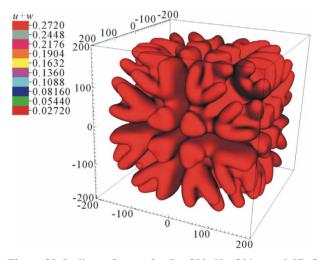


Figure 20. Isolines of u + w for L = 200, N = 201, $v_0 = 0.07$, d = 0.1 and t = 6000.

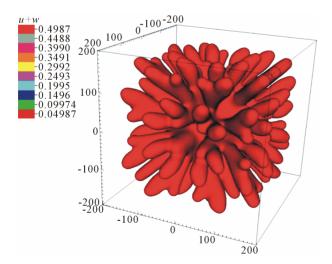


Figure 18. Isolines of u + w for L = 200, N = 201, $v_0 = 0.055$, d = 0.1 and t = 13,000.

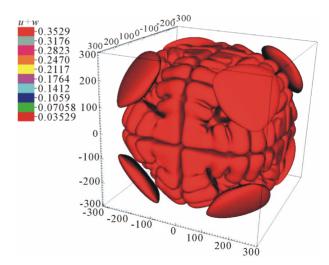


Figure 21. Isolines of u + w for L = 300, N = 301, $v_0 = 0.1$, d = 0.05 and t = 6000.

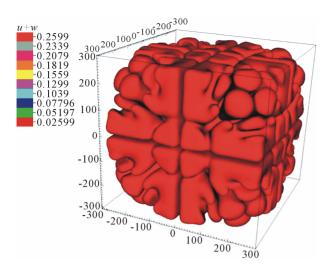


Figure 22. Isolines of u + w for L = 300, N = 301, $v_0 = 0.071$, d = 0.12 and t = 6000.

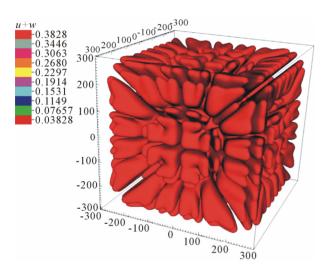


Figure 23. Isolines of u + w for L = 300, N = 301, $v_0 = 0.087$, d = 0.05 and t = 12,000.

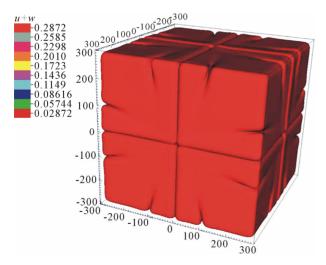


Figure 24. Isolines of u + w for L = 300, N = 301, $v_0 = 0.09$, d = 0.1 and t = 6000.

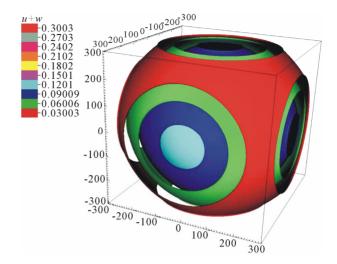


Figure 25. Isolines of u + w for L = 300, N = 301, $v_0 = 0.25$, d = 0.25 and t = 800.

7. Acknowledgements

Partial support of the projects "Applied Mathematics in Physics and Technical Sciences" number MSM684077 0010 of the Ministry of Education, Youth and Sports of the Czech Republic and "Advanced Supercomputing Methods for Implementation of Mathematical Models" number SGS11/161/OHK4/3T/14 is acknowledged.

REFERENCES

- [1] J. Wakita, H. Shimada, H. Itoh, T. Matsuyama and M. Matsushita, "Periodic Colony Formation by Bacterial Species Bacillus Subtilis," *Journal of the Physical Society of Japan*, Vol. 70, No. 3, 2001, pp. 911-919. doi:10.1143/JPSJ.70.911
- [2] M. Mimura, H. Sakaguchi and M. Matsushita, "Reaction-Diffusion Modelling of Bacterial Colony Patterns," *Physica A*, Vol. 282, No. 1-2, 2000, pp. 283-303. doi:10.1016/S0378-4371(00)00085-6
- [3] J. D. Murray, "Mathematical Biology," 3rd Edition, Springer, Berlin, 2002.
- [4] T. Vicsek, "Pattern Formation in Diffusion-Limited Aggregation," *Physical Review Letters*, Vol. 53, No. 24, 1984, pp. 2281-2284. doi:10.1103/PhysRevLett.53.2281
- [5] I. Golding, Y. Kozlovsky, I. Cohen and E. Ben-Jacob, "Studies of Bacterial Branching Growth Using Reaction-Diffusion Models for Colonial Development," *Physica A*, Vol. 260, No. 3-4, 1998, pp. 510-554. doi:10.1016/S0378-4371(98)00345-8
- [6] Y. Kozlovsky, I. Cohen, I. Golding and E. Ben-Jacob, "Lubricating Bacteria Model for Branching Growth of Bacterial Colonies," *Physical Review E*, Vol. 59, No. 6, 1999, pp. 7025-7035. doi:10.1103/PhysRevE.59.7025
- [7] R. Straka and Z. Čulík, "Numerical Aspects of a Bacteria Growth Model," *Proceedings of Algoritmy*, Vysoké Tatry-Podbanské, 13-18 March 2005, pp. 175-184.
- [8] E. Feireisl, D. Hilhorst, M. Mimura and R. Weidenfeld,

- "On a Nonlinear Diffusion System with Resource-Consumer Interaction," *Hiroshima Mathematical Journal*, Vol. 33, No. 2, 2003, pp. 253-295.
- [9] S. Brenner and R. Scott, "The Mathematical Theory of Finite Element Methods," 3rd Edition, Springer, New York, 2008. doi:10.1007/978-0-387-75934-0
- [10] H. Fujita, N. Saito and T. Suzuki, "Operator Theory and Numerical Methods," Elsevier, Amsterdam, 2001.
- [11] V. Thomée, "Galerkin Finite Element Methods for Parabolic Problems," Springer, Berlin, 1997. doi:10.1007/978-3-662-03359-3
- [12] M. Beneš, "Mathematical Analysis of Phase-Field Equations with Numerically Efficient Coupling Terms," *Interfaces and Free Boundaries*, Vol. 3, No. 2, 2001, pp. 201-221. doi:10.4171/IFB/38
- [13] T. Oberhuber, "Finite Difference Scheme for the Will-

- more Flow of Graphs," *Kybernetika*, Vol. 43, No. 6, 2007, pp. 855-867.
- [14] J. Šembera and M. Beneš, "Nonlinear Galerkin Method for Reaction-Diffusion Systems Admitting Invariant Regions," *Journal of Computational and Applied Mathematics*, Vol. 136, No. 1-2, 2001, pp. 163-176. doi:10.1016/S0377-0427(00)00582-3
- [15] W. E. Schiesser, "The Numerical Method of Lines," Academic, New York, 1991.
- [16] G. Dziuk, "Convergence of a Semi-Discrete Scheme for the Curve Shortening Flow," *Mathematical Models and Methods in Applied Sciences*, Vol. 4. No. 4, 1994, pp. 589-606. doi:10.1142/S0218202594000339
- [17] O. Pártl, "Reaction-Diffusion Systems in Mathematical Biology," Master Thesis, Czech Technical University in Prague, Prague, 2012.