Fast and Stable Schemes for Phase Fields Models

M. Brachet*

J.-P. Chehab[†]

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Abstract

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^{*}Institut Elie Cartan de Lorraine, Université de Lorraine, Site de Metz, Bât. A Ile du Saucy, F-57045 Metz Cedex 1, matthieu.brachet@math.univ-metz.fr

[†]Laboratoire Amienois de Mathématiques Fondamentales et Appliquées (LAMFA), UMR 7352, Université de Picardie Jules Verne, 33 rue Saint Leu, 80039 Amiens France , (Jean-Paul.Chehab@u-picardie.fr).

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1 Introduction

Phase fields equations, such as Allen-Cahn's or Cahn-Hilliard's, play an important role in applications since they allow to model natural phenomena; let us cite [1, 2, 16, 26] in material science, [5, 6, 7, 18, 20, 22, 23] in image processing, [22, 25] in chemistry or [10, 19] in ecology and in medicine, just to cite but a few, the list being non-exhaustive. There is also an important interest for these models in the mathematical analysis point of view, see [14, 15, 28]. The simulation of Phase fields models is then an important issue.

The numerical integration of such reaction-diffusion equations can be a delicate task: it needs to recover at the discrete level intrinsic properties of the solution (Energy diminishing, maximum principle) and the presence of small parameter (typically, the interphase length) can generate practical difficulties in the iterations processes with a hard time step restriction, even for fully-implicit schemes; this is due on the way the fixed points problems are solved at each iteration.

The construction of a robust (stable) and efficient (fast) scheme lies on the balance between the advantages and the drawbacks of implicit (stable but costly) and of explicit (fast but with often stability condition) times-marching schemes. For instance, the simple Forward Euler's, can be used only for small time steps; this restriction can be very important, e.g., when considering heat-equation the basic linear part of reaction-diffusion equations. This restriction allows to prevent the expansion of high mode components, the ones that lead to the divergence of the scheme. A way to enhance the stability region is to introduce an approximation to an unconditionally stable scheme. Consider, e.g., Backward Euler's applied to the discretized Heat equation:

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + Au^{(k+1)} = 0, (1)$$

where A is the stiffness matrix, $\Delta t > 0$, the time step; here $u^{(k)}$ is the approximation of the solution at time $t = k\Delta t$ in the spatial approximation space. To simplify the linear system that must be solved at each step, one replaces $Au^{(k+1)}$ by $\tau B(u^{(k+1)} - u^{(k)}) + Au^{(k)}$, where $\tau \geq 0$

and where B is a pre-conditioner of A.

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \tau B(u^{(k+1)} - u^{(k)}) + Au^{(k)} = 0.$$
 (2)

This stabilization procedure, also called RSS scheme (Residual Smoothing Scheme), was introduced independently by [3] and [11] (in the multilevel case), see also [8] for recent developments. It allows to take large time steps while simplifying the linear problem to solve at each step: in that way the stability is enhanced and at the same a save of computation time can be obtained as respect to the classical backward Euler's scheme. Of course the stabilization procedure can be applied to a large variety of schemes that are used for reaction diffusion equations:

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \tau B(u^{(k+1)} - u^{(k)}) + Au^{(k)} + f(u^{(k)}) = 0,$$
(3)

that corresponds to stabilized semi-implicit Euler scheme for, e.g., Allen-Cahn equations, and

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \tau B(\mu^{(k+1)} - \mu^{(k)}) + A\mu^{(k)} = 0,$$

$$\mu^{(k+1)} = \tau B(u^{(k+1)} - u^{(k)}) + Au^{(k)} + f(u^{(k)}) = 0,$$
(5)

$$\mu^{(k+1)} = \tau B(u^{(k+1)} - u^{(k)}) + Au^{(k)} + f(u^{(k)}) = 0, \tag{5}$$

which can be considered for high order or coupled problems such as Cahn-Hilliard's. It must be noted that this stabilization procedure allows to recover the same steady states as the original scheme, this s an important property when considering, e.g., inpainting or image segmentation problems.

The aim of this article is to propose and analyse fast finite differences schemes for phase fields models such as Allen-Cahn's or Cahn-Hilliard's equations, when the space discretization is realized with finite differences compact schemes. The new methods combine high order compact finite difference scheme for the discretization in space together with a stabilization of explicit time schemes implemented by using low coast pre-conditioners of the linear term.

The article is organized as follows: in Section 2 we consider the linear case, we recall the principle of the stabilization (RRS- scheme) and derive stability results for a number of time schemes that will be used in the non linear case. After that, in Section 3, then in Section 4, we introduce and study new stabilized schemes for Allen-Cahn's (then Cahn-Hilliard's) equation. We give in particular conditions to obtain energy diminishing schemes. In Section 5 we present numerical illustrations on pattern dynamics, image segmentation and inpainting.

2 Stabilized schemes in the linear case

We first give here stability results for stabilized-Schemes derived from time marching method in the linear case; these schemes will be used to build new methods for solving nonlinear time dependent problems, as presented in Sections 3 and 4.

Explicit Schemes and stabilization 2.1

Consider the Heat equation

$$\frac{\partial u}{\partial t} - \Delta u = 0, \quad x \in \Omega, \ t > 0,
 u = 0 \quad x \in \partial\Omega, \ t > 0,
 u(x,0) = u_0(x), \qquad x \in \Omega.$$
(6)
(7)

$$u = 0 \quad x \in \partial\Omega, \, t > 0, \tag{7}$$

$$u(x,0) = u_0(x), x \in \Omega. (8)$$

They are several way to express the stability of a scheme, depending on the norm one considers to measure the boundedness of the sequence of time approximations of the solution. We will focus on two following stability notions

• Stability in Energy (a consequence of energy time-diminishing):

$$\sum_{|\alpha|=0}^{m} \aleph_{\alpha} \|D^{\alpha} u(t)\|_{L^{2}\Omega)}^{2} \leq \sum_{|\alpha|=0}^{m} \aleph_{\alpha} \|D^{\alpha} u(t')\|_{L^{2}\Omega}^{2}, \ \forall t > t'.$$

with $\aleph_{\alpha} \geq 0$, $\sum \aleph_{\alpha} > 0$.

• L^{∞} Stability (a consequence of the maximum principle):

$$\exists L > 0/\|u(t)\|_{L^{\infty}} \le L, \forall t \ge 0.$$

The space discretization of (6) leads to the differential system

$$\frac{du}{dt} + Au = 0, \quad x \in \Omega, \ t > 0, \tag{9}$$

$$u(0) = u_0, \tag{10}$$

Here A is the stiffness matrix. The time numerical integration of (9) produces a sequence of vectors $u^k \simeq u(k\Delta t)$, and we define the stability of the schemes as

• Stability in Energy (discrete Energy diminishing), e.g.,

$$\frac{1}{2} < Au^{k+1}, u^{k+1} > \leq \frac{1}{2} < Au^k, u^k >, \forall k$$

• L^{∞} Stability

$$\exists L > 0, /\max_{i} |u_{i}^{k}| \leq L, \forall k.$$

These notions of stability will be used also in the nonlinear case, especially for Allen-Cahn's equation.

Let us first recall a simple but useful result, [8]. Let A and B be two $n \times n$ symmetric positive definite matrices. We assume that there exist two strictly positive constant α and β such that

$$\alpha < Bu, u > \le < Au, u > \le \beta < Bu, u >, \forall u \in \mathbb{R}^n$$
(11)

The simplest stabilized scheme is obtained from Backward Euler's as

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \tau B(u^{(k+1)} - u^{(k)}) + Au^{(k)} = 0, \tag{12}$$

where $\tau \geq 0$; Forward Euler is recovered for $\tau = 0$ while Backward Euler's is obtained for $\tau = 1$ and B = A. We have the stability conditions, see [8],

Proposition 2.1 Assume that A and B are two SPD matrices. Under hypothesis (11), we have the following stability conditions:

- If $\tau \geq \frac{\beta}{2}$, the schemes (18) and (21) are unconditionally stable (i.e. stable $\forall \Delta t > 0$),
- If $\tau < \frac{\beta}{2}$, then the scheme is stable for $0 < \Delta t < \frac{2}{\left(1 \frac{2\tau}{\beta}\right)\rho(A)}$.

Of course, instead of Euler's method, we can consider second order schemes such as Gear's or Crank Nicolson's and apply the same stabilization procedure. We can establish the following result:

Proposition 2.2 Consider the stabilized scheme derived from Gear's method

$$\frac{1}{2\Delta t}(3u^{(k+1)} - 4u^{(k)} + u^{(k-1)}) + \tau B(u^{(k+1)} - u^{(k)}) + Au^k = 0$$

We have the following stability conditions

- If $\tau \geq \frac{\beta}{2}$, then (1) is unconditionally stable
- If $\tau < \frac{\beta}{2}$, then (1) is stable when

$$0 < \Delta t < \frac{2}{\rho(A)(1 - \frac{2\tau}{\beta})}$$

Proof. We start from the identity

$$<3u^{(k+1)}-4u^{(k)}+u^{(k-1)},u^{(k+1)}-u^{(k)}>\\ =2\|u^{(k+1)}-u^{(k)}\|^2+\frac{1}{2}(\|u^{(k+1)}-u^{(k)}\|^2\\ -\|u^{(k)}-u^{(k-1)}\|^2+\|u^{(k+1)}-2u^{(k)}+u^{(k-1)}\|^2)$$

We now take the scalar product of each term of (1) with $u^{(k+1)} - u^{(k)}$ and obtain

$$\begin{split} &2\|u^{(k+1)}-u^{(k)}\|^2+\frac{1}{2}(\|u^{(k+1)}-u^{(k)}\|^2-\|u^{(k)}-u^{(k-1)}\|^2+\|u^{(k+1)}-2u^{(k)}+u^{(k-1)}\|^2)\\ &+2\Delta t\left(\tau < B(u^{(k+1)}-u^{(k)}),u^{(k+1)}-u^{(k)}>+< Au^{(k)},u^{(k+1)}-u^{(k)}>\right)=0 \end{split}$$

Using the parallelogram identity on the last term, we find

$$\begin{split} &2\|u^{(k+1)}-u^{(k)}\|^2+\frac{1}{2}(\|u^{(k+1)}-u^{(k)}\|^2-\|u^{(k)}-u^{(k-1)}\|^2+\|u^{(k+1)}-2u^{(k)}+u^{(k-1)}\|^2)\\ &+2\Delta t\left(\tau < B(u^{(k+1)}-u^{(k)}),u^{(k+1)}-u^{(k)}>\\ &+\frac{1}{2}(< Au^{(k+1)},u^{(k+1)}>-< Au^{(k)},u^{(k)}>-< A(u^{(k+1)}-u^{(k)}),u^{(k+1)}-u^{(k)}>=0 \end{split}$$

Now, we let $E^{k+1} = \frac{1}{2} < Au^{(k+1)}, u^{(k+1)} > +\frac{1}{2} ||u^{(k+1)} - u^{(k)}||^2$ and we obtain,

$$2\|u^{(k+1)} - u^{(k)}\|^2 + \frac{1}{2}\|u^{(k+1)} - 2u^{(k)} + u^{(k-1)}\|^2 + 2\Delta t(E^{k+1} - E^k) + 2\Delta t(\tau < B(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} > -\frac{1}{2} < A(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} >)$$

The stability is obtained when $E^{k+1} < E^k$, hence the conditions. The classical Gear's method is obtained taking $\tau = 0$.

The stabilized Gear's scheme can be implemented as follows:

Algorithm 1 : Stabilized-Gear

1: for $k = 0, 1, \cdots$ until convergence do

Solve $(3.Id + \tau 2\Delta tB)\delta = -2\Delta t(u^{(k)} - u^{(k-1)} + Au^{(k)})$

Set $u^{(k+1)} = u^{(k)} + \delta$

4: end for

Consider now the stabilization of Crank-Nicolson scheme:

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \frac{1}{2} (Au^{(k+1)} + Au^{(k)}) = 0, \tag{13}$$

(14)

This scheme can be rewritten as

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \frac{1}{2} (Au^{(k+1)} - Au^{(k)}) = -Au^{(k)}, \tag{15}$$

(16)

and implemented following the two steps:

- Solve $(Id + \frac{\Delta t}{2}A)\delta = -\Delta t A u^{(k)}$
- Set $u^{(k+1)} = u^{(k)} + \delta$

The stabilization scheme consists in replacing the implicit part by a simplified one using a preconditioning matrix B of A, namely

- Solve $(Id + \tau \frac{\Delta t}{2}B)\delta = -\Delta t A u^{(k)}$
- Set $u^{(k+1)} = u^{(k)} + \delta$

The implementation of the stabilzed-simplified CN scheme reads as

Algorithm 2: Stabilized Crank-Nicolson

- 1: **for** $k = 0, 1, \cdots$ until convergence **do**
- Solve $(Id + \frac{\tau \Delta t}{2}B)\delta = -\Delta tAu^{(k)}$ Set $u^{(k+1)} = u^{(k)} + \delta$ 2:
- 4: end for

This scheme is consistent with the classical CN one since, for $\tau = 1$ and B = A the original method is recovered. We can easily derive the following stability result for this scheme:

Proposition 2.3 Assume that A and B are two SPD matrices. Under hypothesis (11), we have the following stability conditions:

- If $\tau \geq \beta$, the schemes (18) and (21) are unconditionally stable (i.e. stable $\forall \Delta t > 0$)
- If $\tau < \beta$, then the scheme is stable for $0 < \Delta t < \frac{2}{\left(1 \frac{\tau}{\beta}\right)\rho(A)}$.

Proof. It suffices to replace τ by $\frac{\tau}{2}$ in Proposition (2.1).

2.2 Discrete Maximum Principle

Another important stability property, crucial in a number of applications, is the L^{∞} stability guaranted by a maximum principle. We have the

Proposition 2.4 Assume that $f \ge 0$ and $u^{(0)} \ge 0$. Assume in addition that $Id + \Delta tB$ is a M-matrix for all $\Delta t > 0$. Set $D = \tau B - A$ and $I = \{i \in \{1, \dots, n\}/D_{ii} < 0\}$. If

$$D_{i,j} \ge 0, \forall i, j = 1, \dots, N, i \ne j \text{ and } 0 < \Delta t < \frac{1}{\max_{i \in I} |D_{ii}|}, i \in I,$$

then $u^{(k)} \geq 0, \forall k$.

Proof. Let k be fixed and assume that $u^{(k)} \geq 0$. We have

$$(Id + \tau \Delta tB)u^{(k+1)} = (Id + \Delta t(\tau B - A))u^{(k)} + \delta tf$$

The matrix $Id + \tau \Delta tB$ is a M-matrix since B is also one, as a direct consequence. Hence, a sufficient condition to have $u^{(k+1)} \geq 0$ is $(Id + \Delta t(\tau B - A))u^{(k)} \geq 0$, this is guaranted when the matrix $R = Id + \Delta t(\tau B - A) = Id + \Delta tD$ has all positive entries, say

$$\begin{cases} 1 + \Delta t(\tau B_{ii} - A_{ii}) \ge 0 & i = 1, \dots, n \\ \tau B_{ij} - A_{ij} \ge 0 & i = 1, \dots, n \ i \ne j \end{cases}$$

Hence the result by a simple induction. ■

It is usual that the discrete Maximum principle is satisfied for small values of Δt ; we recover particularly the conditions that must be satisfied for the Crank-Nicolson scheme taking $\tau = \frac{1}{2}$ and A = B, see [24].

2.3 ADI Stabilized Scheme

A important issue for a fast simulation of parabolic equations is the use of splitting methods. We give here stabilized versions of classical ADI schemes. Consider the linear differential system

$$\frac{dU}{dt} + AU = 0$$

with $A = A_1 + A_2$. Let B_1 and B_2 be pre-conditioners of A_1 and A_2 respectively and τ_1 , τ_2 two positive real numbers. All the matrices are supposed to be symmetric positive definite. We introduce the stabilized ADI-schemes

$$\frac{u^{(k+1/2)} - u^{(k)}}{\Delta t} + \tau_1 B_1 (u^{(k+1/2)} - u^{(k)}) = -A_1 u^{(k)}, \tag{17}$$

$$\frac{u^{(k+1)} - u^{(k+1/2)}}{\Delta t} + \tau_2 B_2(u^{(k+1)} - u^{(k+1/2)}) = -A_2 u^{(k+1/2)},\tag{18}$$

and the Strang's Splitting

$$\frac{u^{(k+1/3)} - u^{(k)}}{\Delta t/2} + \tau_1 B_1(u^{(k+1/3)} - u^{(k)}) = -A_1 u^{(k)}, \tag{19}$$

$$\frac{u^{(k+2/3)} - u^{(k+1/3)}}{\Delta t} + \tau_2 B_2(u^{(k+2/3)} - u^{(k+1/3)}) = -A_2 u^{(k+1/3)}, \tag{20}$$

$$\frac{u^{(k+1)} - u^{(k+2/3)}}{\Delta t/2} + \tau_1 B_1 (u^{(k+1)} - u^{(k+2/3)}) = -A_1 u^{(k+2/3)}, \tag{21}$$

considering $A = \sum_{i=1}^{m} A_i$ and $B = \sum_{i=1}^{m} B_i$ and the splitting

$$\frac{u^{(k+i/m)} - u^{(k+(i-1)/m)}}{\Delta t} + \tau_i B_i (u^{(k+i/m)} - u^{(k+(i-1)/m)}) = -A_i u^{(k+(i-1)/m)}, \tag{22}$$

We recall that

As a direct consequence of proposition 2.1, we can prove the following result

Proposition 2.5 Under hypothesis (11), we have the following stability conditions:

- If $\tau_i \geq \frac{\beta_i}{2}$, i = 1, 2 the scheme (18) is unconditionally stable (i.e. stable $\forall \Delta t > 0$)
- If $\tau_i < \frac{\beta_i}{2}$, i = 1, 2, then the scheme is stable for

$$0 < \Delta t < Min(\frac{2}{\left(1 - \frac{2\tau_1}{\beta_1}\right)\rho(A_1)}, \frac{2}{\left(1 - \frac{2\tau_2}{\beta_2}\right)\rho(A_2)}).$$

Proof. It suffices to apply proposition 2.1 to each system.

Numerical illustrations

Before presenting the stabilized schemes for phase fields models, we give hereafter some numerical illustrations on linear problems when discretized in space by finite differences compact schemes, focusing on Neumann boundary conditions. We propose as in [8] to use a (lower) second order discretization matrix for preconditioning the underlining matrices. We first consider the Elliptic Neumann problem then the Heat equation.

Finite difference Preconditioning for compact schemes and the Neumann problem

We use finite differences compact schemes for the discretization in space: in two words these schemes are nonlocal (they have an implicit part), and they allows to reach an accuracy comparable to the spectral one; we refer to the classical book of Collatz [12] and to the seminal paper of Lele [21]; the details of the building of Poisson problems with Homogeneous Dirichlet or Neumann boundary conditions can be found also in [8].

The 2D and 3D reaction-diffusion problems we will simulate (Allen-Cahn's or Cahn-Hilliard's equations) are completed with Neumann Boundary Conditions, so we start by considering teh basic linear problem:

$$\alpha u - \Delta u = f \quad \text{in } \Omega =]0, 1[^{2,3}, \tag{23}$$

$$\alpha u - \Delta u = f \quad \text{in } \Omega =]0, 1[^{2,3},$$

$$\frac{\partial u}{\partial n} = 0 \quad \text{on } \partial\Omega,$$
(23)

when discretized by fourth order compact schemes. The preconditioning matrix corresponds to the usual five point scheme finite difference schemes. A random r.h.s b=1-2*rand(N,1) is chosen so in order to have the presence of a late number of frequencies. The initial data is u=0, the tolerance parameter 10^{-10} , we took $\alpha=\beta=1$ but the results are very close for other values of these parameters. The result we report is the maximum number of external iterations to convergence, on 5 independent numerical resolutions, the number of discretization point per direction n is reported as (n). The stiffness matrices are of respective sizes $n^2 \times n^2$ (2D problem) and $n^3 \times n^3$ (3D problem). The preconditioning systems are solved using the cosine FFT.

Problem	# it. (n)	# it. (n))	# it. (n)	# it. (n)	#it. (n)	#it. (n)
Poisson 2D	16 (n=15)	15 (n=31)	14 (n=63)	13 (n=127)	12 (n=255)	(n=511)
Poisson 3D	30 (n=15)	30 (n=31)	(n=63)			

Table 1: Solutions of 2D and 3D Neumann problem with GMRES and second order preconditioner

Of course, due the implicity of the scheme, the fourth order discretization matrix A of $-\Delta$ is nonsymmetric while the preconditioning matrix B is. However, in practice the RSS method is still efficient. This is due to the small size of the skew-symmetric part of A, see [8].

2.4.2 Heat Equation

For the 2D and the 3D heat equation, we here compare different RSS schemes presented above (ADI, Gear's, CN and Euler's) in order to illustrate the stability results and to point out the owl of the stabilization parameter τ .

USE THE MATLAB CODES IN THE DIRECTORY ADI_MATLAB. see program chaleur_2D_splitting.m

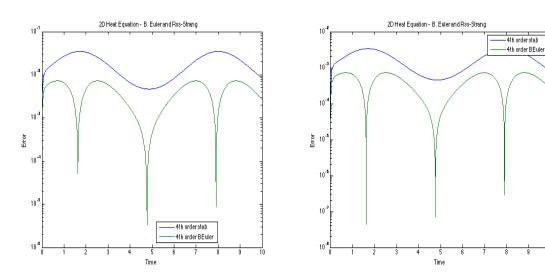


Figure 1: Solution of the heat equation with $\Delta t = 0.01$, (left) and $\Delta t = 0.001$ (right) n = 31, $\tau = 1$

The error is clearly in Δt .

We now consider the 3D Heat equation, with Homogeneous Neumann BC. The error is clearly in Δt .

- Solution of the Neumann Problem with Preconditioned GMRES solve Neum 3D CS.m
- Neumann Problem test dct neumann.m
- Heat equation test chaleur.m

•

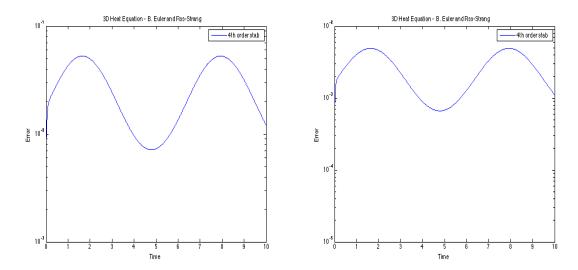


Figure 2: Solution of the 3D heat equation with $\Delta t = 0.01$, (left) and $\Delta t = 0.001$ (right) n = 31, $\tau = 1$

3 Allen-Cahn's equation

Let $\Omega \subset \mathbb{R}^n$, n=2,3 a regular open bounded set. We here consider the simple Allen-Cahn equation

$$\frac{\partial u}{\partial t} + M(-\Delta u + \frac{1}{\epsilon^2} f(u)) = 0, \quad x \in \Omega, t > 0$$
 (25)

$$\frac{\partial u}{\partial n} = 0 \qquad t > 0, \tag{26}$$

$$u(0, x) = u_0(x), \qquad x \in \Omega. \tag{27}$$

$$u(0,x) = u_0(x), \qquad x \in \Omega. \tag{27}$$

which describes the process of phase separation in iron alloys [1, 2], including order-disorder transitions: M is the mobilty (taken to be 1 for simplicity), $F = \int_{-\infty}^{u} f(v)dv$ is the free energy, u is the (non-conserved) order parameter, ϵ is the interface length. The homogenous Neumann boundary condition implies that there is not a loss of mass outside the domain Ω . It is important to note that here is a competition between the potential term and the diffusion term: regularization in phase transition. Two important properties are satisfied by the solution and must be captured by the numerical scheme (intrinsically or numerically):

• the energy diminishing: Allen-Cahn equation is a gradient flow for the energy

$$E(u) = \frac{1}{2} \int_{\Omega} \|\nabla u\|^2 dx + \frac{1}{\epsilon^2} F(u) dx,$$

so
$$E(u(t)) \leq E(u(t')), \forall t \geq t'$$
,

• the maximum principle: $|u(.,t)|_{L^{\infty}} \leq L, \forall t > 0.$

Energy diminishing schemes

We set $E(u) = \frac{1}{2} \langle Au, u \rangle + \frac{1}{\epsilon^2} \langle F(u), \mathbf{1} \rangle$, where F is a primitive of f that we choose such that F(0) = 0; 1 is the vector whose components are all equal to 1. We say that the scheme is energy decreasing if

$$E(u^{(k+1)}) < E(u^{(k)}).$$

We here first recall somme schemes and their stability conditions, see also [8].

The semi-implicit Scheme applied to the pattern evolution A-C equation:

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + Au^{(k+1)} + \frac{1}{\epsilon^2} f(u^{(k)}) = 0$$
 (28)

is energy diminishing under the time step restriction

$$0 < \Delta t < \frac{\epsilon^2 L}{2},$$

where $|f'|_{\infty} \le L$, see [27].

The stabilization of its linear part is

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \tau B(u^{(k+1)} - u^{(k)}) + \frac{1}{\epsilon^2} f(u^{(k)}) = -Au^{(k)}.$$
 (29)

We have the stability result, [8]:

Theorem 3.1 Assume that f is C^1 and $|f'|_{\infty} \leq L$. We have the following stability conditions

• If
$$\tau \geq \frac{\beta}{2}$$
 then
$$-if\left(\frac{\tau}{\beta} - \frac{1}{2}\right)\lambda_{min} - \frac{L}{2\epsilon^2} \geq 0 \text{ then the scheme is unconditionally stable,}$$

$$-if\left(\frac{\tau}{\beta} - \frac{1}{2}\right)\lambda_{min} - \frac{L}{2\epsilon^2} < 0 \text{ then the scheme is stable for}$$

$$0 < \Delta t < \frac{1}{\frac{L}{2\epsilon^2} - \left(\frac{\tau}{\beta} - \frac{1}{2}\right)\lambda_{min}},$$

• If $\tau < \frac{\beta}{2}$ then the scheme is stable for

$$0 < \Delta t < \frac{1}{\frac{L}{2\epsilon^2} - \left(\frac{\tau}{\beta} - \frac{1}{2}\right)\rho(A)}.$$

A more stable way to overcome the stability restriction is to consider directly AC equation as a gradient system with a natural diminishing energy property. A first unconditionally stable scheme is ([14, 15])

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + Au^{(k+1)} + \frac{1}{\epsilon^2} DF(u^{(k)}, u^{(k+1)}) = 0, \tag{30}$$

where

$$DF(u,v) = \begin{cases} \frac{F(u) - F(v)}{u - v} & \text{if } u \neq v, \\ f(u) & \text{if } u = v. \end{cases}$$

In [8] it was introduced the RSS-scheme

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \tau B(u^{(k+1)} - u^{(k)}) + DF(u^{(k+1)}, u^{(k)}) = -Au^{(k)}$$
(31)

which enjoys of the following stability condition, see [8] for the proof.

Proposition 3.2 Under hypothesis \mathcal{H}

• if $\tau \geq \frac{\beta}{2}$, the RSS scheme is unconditionally stable,

• if $\tau < \frac{\beta}{2}$, the RSS scheme is stable under condition

$$0 < \Delta t < \frac{\beta}{\rho(A)(\frac{\beta}{2} - \tau)}.$$

Finally, unconditionally stable scheme is to use the so-called convex splitting, [17, 13]. These schemes are based on a proper splitting of the free energy term

$$F(u) = F_c(u) - F_e(u),$$

where $F_* \in \mathcal{C}^2(\mathbb{R}^n, \mathbb{R}), * = c \text{ or } * = e.$

ii. F_* is strictly convex in \mathbb{R}^n , *=c or *=e.

iii.
$$\langle [\nabla F_e(u)]u, u \rangle \geq -\lambda, \forall u \in \mathbb{R}^n.$$

The scheme reads as

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + Au^{(k+1)} + \nabla F_c(u^{(k+1)}) = +\nabla F_e(u^{(k)}), \tag{32}$$

Its sabilization reads as

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \tau B(u^{(k+1)} - u^{(k)}) + \nabla F_c(u^{(k+1)}) = -Au^{(k)} + \nabla F_e(u^{(k)}), \tag{33}$$

and we can prove the

Theorem 3.3 • If $(\tau\beta - \frac{1}{2}\rho(A) + (\hat{\lambda} - |\lambda|) > 0$ then the scheme is unconditionally stable

• Else it is stable under condition

$$0 < \Delta t < \frac{1}{(\frac{1}{2} - \tau \beta)\rho(A) + |\lambda| - \hat{\lambda}}.$$

3.2 Splitting schemes

We follow [22] who proposed for the so-called double well potential case the scheme $(F(u) = \frac{1}{4}(1-u^2)^2)$ the splitting scheme

$$\frac{u^* - u^{(k)}}{\Delta t} + \frac{1}{2}(Au^* + Au^{(k)}) = 0, (34)$$

$$\frac{u^{(k+1)} - u^*}{\Delta t} = \frac{u^{(k+1)} - (u^{(k+1)})^3}{\epsilon^2}.$$
 (35)

The last equation can be simplified since it correspond to a one-step approximation by backward Euler's to the differential equation

$$\frac{du}{dt} = \frac{u - u^3}{\epsilon^3},\tag{36}$$

whose the solution is

$$u(t) = \frac{u(0)}{\sqrt{e^{-2\frac{t}{\epsilon^2}} + u(0)^2(1 - e^{-2\frac{t}{\epsilon^2}})}}$$

Hence the simplified scheme

$$\frac{u^* - u^{(k)}}{\Delta t} + \frac{1}{2}(Au^* + Au^{(k)}) = 0, (37)$$

$$u^{(k+1)} = \frac{u^*}{\sqrt{e^{-2\frac{\Delta t}{e^2}} + (u^*)^2 (1 - e^{-2\frac{\Delta t}{e^2}})}}$$
(38)

is rather considered and have nice stability properties, see [22].

We get give here a first stability result that will be useful to prove

Theorem 3.4 Let $\mathbf{1} = (1, 1, \dots, 1)^T \in \mathbb{R}^N$. Assume that $A\mathbf{1} = 0$ and that $Id + \Delta tA$ enjoys of the discrete maximum principle. Assume that $|u_i^{(0)}| \leq 1, i = 1, \dots, N$. Then the sequence $u^{(k)}$ defined by

$$\frac{u^* - u^{(k)}}{\Delta t} + Au^* = 0, (39)$$

$$\frac{u^* - u^{(k)}}{\Delta t} + Au^* = 0,$$

$$u^{(k+1)} = \frac{u^*}{\sqrt{e^{-2\frac{\Delta t}{\epsilon^2}} + (u^*)^2 (1 - e^{-2\frac{\Delta t}{\epsilon^2}})}}$$
(40)

satisfies $|u_i^{(k)}| \le 1, i = 1, \dots N$

Proof. We proceed by induction. First of all we show that if $|u^{(k)}| \leq 1$ then $|u^{(*)}| \leq 1$. We have

$$(Id + \Delta tA)(u^* - \mathbf{1}) = (u^{(k)} - \mathbf{1}).$$

Hence, by the maximum principle, if $u^{(k)} - 1 \le 0$ then $u^{(*)} - 1 \le 0$. Replacing $(u^{(k)} - 1)$ (resp. $(u^* - 1)$ by $(u^{(k)} + 1)$ (resp. $(u^* + 1)$) we find that $u^{(*)} + 1 \ge 0$ and conclude that $-1 \le u^* \le 1$. Now, to conclude, it suffices to show that

$$\left| \frac{x}{\sqrt{e^{-2\frac{\Delta t}{\epsilon^2}} + (x)^2 (1 - e^{-2\frac{\Delta t}{\epsilon^2}})}} \right| \le 1 \ \forall x \in [-1, 1], \ \forall \Delta t > 0, \ \forall \epsilon^2 > 0.$$

We set for convenience $\gamma = e^{-2\frac{\Delta t}{\epsilon^2}} \in [0,1]$. We start from

$$x^2 \le 1 \iff \gamma x^2 \le \gamma \iff x^2 \le (1 - \gamma)x^2 + \gamma \iff \frac{x^2}{(1 - \gamma)x^2 + \gamma} \le 1$$

The result is obtained by taking the square-root of this last expression.

We can define a stabilized version of this splitting scheme as

$$\frac{u^* - u^{(k)}}{\Delta t} + \tau B(u^* - u^{(k)}) = -Au^{(k)},\tag{41}$$

$$u^{(k+1)} = \frac{u^*}{\sqrt{e^{-2\frac{\Delta t}{\epsilon^2}} + (u^*)^2 (1 - e^{-2\frac{\Delta t}{\epsilon^2}})}}$$
(42)

To implement RRS-like version of this splitting scheme it then suffices to replace the first step by a RSS-CN scheme as proposed in section 2. We then obtain the RSS-splitting scheme

Algorithm 3: RSS splitting for Allen Cahn

1: **for**
$$k = 0, 1, \dots$$
do

2: Solve
$$(Id + \frac{\tau \Delta t}{2} \epsilon B)\delta = -\Delta t A u^{(k)}$$

3: **Set**
$$u^{(*)} = u^{(k)} + \delta$$

1: Iof
$$k = 0, 1, \cdots$$
 do
2: Solve $(Id + \frac{\tau \Delta t}{2} \epsilon B)\delta = -\Delta t A u^{(k)}$
3: Set $u^{(*)} = u^{(k)} + \delta$
4: Set $u^{(k+1)} = \frac{u^*}{\sqrt{e^{-2\frac{\Delta t}{\epsilon^2}} + (u^*)^2(1 - e^{-2\frac{\Delta t}{\epsilon^2}})}}$

5: end for

SEE Program RSSsplittingAC3DCS.m.

The proof for the L^{∞} -stability of the classical θ scheme with a second order FD matrix is given in ([24]) p-33, but is based on a pointwize analysis.

Theorem 3.5 We make the assumptions of Proposition 2.4 on A and B. Let $\mathbf{1} = (1, 1, \dots, 1)^T \in \mathbb{R}^N$. Assume that $A\mathbf{1} = 0$, $B\mathbf{1} = 0$ and that $|u_i^{(0)}| \leq 1, i = 1, \dots N$. Then the sequence $u^{(k)}$ defined by (41)-(42) satisfies $|u_i^{(k)}| \leq 1, i = 1, \dots N$

Proof. According to the proof of Theorem 3.4, it suffices to show that if $|u^{(k)}| \leq 1$ then $|u^{(*)}| \leq 1$. This is automatically provided using Proposition 2.4.

4 Cahn-Hilliard equation

We here present briefly Cahn-Hilliard (CH) equations used for Phase transition and for image inpainting. We introduce new stabilized schemes and establish stability properties.

4.1 The models

4.1.1 Cahn-Hilliard and Patterns dynamics

The CH equation describes the process of phase separation, by which the two components of a binary fluid spontaneously separate and form domains pure in each component. It writes as

$$\frac{\partial u}{\partial t} - \Delta(-\Delta u + \frac{1}{\epsilon^2}f(u)) = 0, \tag{43}$$

$$\frac{\partial u}{\partial n} = 0,\tag{44}$$

$$\frac{\partial}{\partial n} \left(\Delta u - \frac{1}{\epsilon^2} f(u) \right) = 0, \tag{45}$$

$$u(0,x) = u_0(x) \tag{46}$$

This equation enjoys of the following properties

- \bullet Conservation of the mass: $\bar{u}=\int_{\Omega}u(x,t)dx=\int_{\Omega}u_{0}(x)dx$
- Decay of the energy in time

$$\frac{\partial E(u)}{\partial t} = -\int_{\Omega} |\nabla(-\Delta u + \frac{1}{\epsilon^2} f(u))|^2 dx \le 0$$

A classical way to study and to simulate CH is to decouple the equation as follows:

$$\frac{\partial u}{\partial t} - \Delta \mu = 0, \quad \text{in } \Omega, t > 0 \tag{47}$$

$$\mu = -\Delta u + \frac{1}{\epsilon^2} f(u), \quad \text{in } \Omega, tt > 0$$
(48)

$$\frac{\partial u}{\partial n} = 0, \frac{\partial \mu}{\partial n} = 0, \quad \text{on } \partial \Omega, t > 0$$
 (49)

$$u(0,x) = u_0(x) \qquad \text{in } \Omega \tag{50}$$

4.1.2 The inpainting problem

Cahn-Hilliard equations allow here to in paint a tagged picture. Let g be the original image and $D \subset \Omega$ the region of Ω in which the image is deterred. The idea is to add a penalty term that forces the image to remain unchanged in $\Omega \setminus D$ and to reconnect the fields of g inside D. Let $\lambda >> 1$

$$\frac{\partial u}{\partial t} - \Delta(-\epsilon \Delta u + \frac{1}{\epsilon} f(u)) + \lambda \chi_{\Omega \setminus D}(x)(u - g) = 0, \tag{51}$$

$$\frac{\partial u}{\partial n} = 0 \quad \frac{\partial}{\partial n} \left(\Delta u - \frac{1}{\epsilon^2} f(u) \right) = 0, \tag{53}$$

$$u(0,x) = u_0(x) \tag{54}$$

Here $\chi_{\Omega \backslash D}(x) = \left\{ \begin{array}{ll} 1 & \text{if } x \in \Omega \backslash D, \\ 0 & \text{else} \end{array} \right.$

- The presence of the penalization term $\lambda \chi_{\Omega \setminus D}(x)(u-g)$ forces the solution to be close to g in $\Omega \setminus D$ when $\lambda >> 1$
- \bullet The Cahn-Hilliard flow has as effect to connect the fields inside D
- here ϵ will play the role of the "contrast". A post-processing is possible using a thresholding procedure.

4.2 The Stabilized Scheme

The semi-implicit scheme

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + A\mu^{(k+1)} = 0, (55)$$

$$\mu^{(k+1)} = \epsilon A u^{(k+1)} + \frac{1}{\epsilon} f(u^{(k)}), \tag{56}$$

suffers from a hard time step restriction, its energy stability is guaranteed for

$$0 < \Delta t < \epsilon^2$$

see [27] We derive the Stabilized-Scheme from the backward Euler's (55)-(56) by replacing $Az^{(k+1)}$ by $\tau B(z^{(k+1)}-z^{(k)})+Az^{(k)}$ for z=u or $z=\mu$. We obtain

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \tau B(\mu^{(k+1)} - \mu^{(k)}) + A\mu^{(k)} = 0, \tag{57}$$

$$\mu^{(k+1)} = \epsilon \tau B(u^{(k+1)} - u^{(k)}) + \epsilon A u^{(k)} + \frac{1}{\epsilon} f(u^{(k)}).$$
 (58)

We remark that this scheme preserves the steady state. We now address a stability analysis.

Theorem 4.1 We have the following stability conditions in the linear and in the nonlinear case:

- Linear case $f \equiv 0$: If $\tau > \beta$, then the scheme (57)-(58) is unconditionally stable.
- Nonlinear case: Assume that there exists $\theta \in]0,1[$ such that

$$-\tau(1-\theta) \ge \beta$$
$$-\tau > (1+\beta)$$

Then, the stability is obtained when

$$-0 < \Delta t < \frac{4\epsilon^2}{L^2}$$

$$-0 < \Delta t < \frac{\epsilon(\tau - \beta/2)}{\frac{\tau}{2\theta} + \frac{\beta}{2\eta}} \text{ for some } \eta > 0$$

Proof. We first consider the linear case $(f \equiv 0)$.

We take the scalar product of (57) with $\mu^{(k+1)}$ and of (58) with $u^{(k+1)} - u^{(k)}$. After the use of the parallelogram identity and usual simplifications, we obtain, on the one hand

$$\begin{split} &< u^{(k+1)} - u^{(k)}, \mu^{(k+1)} > + \frac{\Delta t \tau}{2} \left(< B \mu^{(k+1)}, \mu^{(k+1)} > - < B \mu^{(k)}, \mu^{(k)} > \right. \\ &+ &< B(\mu^{(k+1)} - \mu^{(k)}), \mu^{(k+1)} - \mu^{(k)} > \\ &+ \frac{\Delta t}{2} \left(< A \mu^{(k+1)}, \mu^{(k+1)} > + < A \mu^{(k)}, \mu^{(k)} > \right. \\ &- &< A(\mu^{(k+1)} - \mu^{(k)}), \mu^{(k+1)} - \mu^{(k)} > = 0, \end{split}$$

and on the other hand

$$< u^{(k+1)} - u^{(k)}, \mu^{(k+1)}> \\ = \tau \epsilon < B(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)}> \\ + \frac{1}{2} \epsilon \left(< Au^{(k+1)}, u^{(k+1)}> - < Au^{(k)}, u^{(k)}> - < A(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)}> \right)$$

Taking the difference of the last two identities, we obtain

$$\begin{split} &\epsilon\{\tau < B(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} > -\frac{1}{2} < A(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} > \} \\ &+ \Delta t \{\frac{\tau}{2} < B(\mu^{(k+1)} - \mu^{(k)}), \mu^{(k+1)} - \mu^{(k)} > -\frac{1}{2} < A(\mu^{(k+1)} - \mu^{(k)}), \mu^{(k+1)} - \mu^{(k)} > \} \\ &+ \frac{\Delta t}{2} \left(\tau < B\mu^{(k)}, \mu^{(k)} > - < A\mu^{(k)}, \mu^{(k)} > \right) \\ &+ \frac{\Delta t}{2} \left(< A\mu^{(k+1)}, \mu^{(k+1)} > + \tau < B\mu^{(k+1)}, \mu^{(k+1)} > \right) + R^{k+1} - R^k = 0, \end{split}$$

where

$$R^{k+1} = \frac{1}{2}\epsilon < Au^{(k+1)}, u^{(k+1)} >$$

The scheme is then stable when $R^{k+1} < R^k$. Now using (11), we obtain

$$\begin{split} & \epsilon \{ \tau < B(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} > -\frac{1}{2} < A(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} > \} \\ & + \Delta t \{ \frac{\tau}{2} < B(\mu^{(k+1)} - \mu^{(k)}), \mu^{(k+1)} - \mu^{(k)} > -\frac{1}{2} < A(\mu^{(k+1)} - \mu^{(k)}), \mu^{(k+1)} - \mu^{(k)} > \} \\ & + \frac{\Delta t}{2} \left(\tau < B\mu^{(k)}, \mu^{(k)} > - < A\mu^{(k)}, \mu^{(k)} > \right) \\ \geq & \\ & \epsilon (\tau - \frac{\beta}{2}) B(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} > \\ & + \frac{\Delta t}{2} (\tau - \beta) < B(\mu^{(k+1)} - \mu^{(k)}), \mu^{(k+1)} - \mu^{(k)} > \\ & + \frac{\Delta t}{2} (\tau - \beta) < B\mu^{(k)}, \mu^{(k)} > \end{split}$$

Hence the sufficient stability conditions.

Consider now the general case $f \neq 0$. First of all, as in [27], we take the Taylor expansion of the term $\frac{1}{\epsilon}f(u^k) = \frac{1}{\epsilon} < F(u^{(k+1))} - F(u^{(k)}), 1 > + \frac{1}{2\epsilon} < f'(\xi^{(k)})(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} > .$ Hence, we deduce from the previous inequality

$$\begin{split} & \epsilon(\tau - \frac{\beta}{2})B(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} > \\ & + \frac{\Delta t}{2}(\tau - \beta) < B(\mu^{(k+1)} - \mu^{(k)}), \mu^{(k+1)} - \mu^{(k)} > \\ & + \frac{\Delta t}{2}(\tau - \beta) < B\mu^{(k)}, \mu^{(k)} > \\ & + \frac{\Delta t}{2}\left(< A\mu^{(k+1)}, \mu^{(k+1)} > + \tau < B\mu^{(k+1)}, \mu^{(k+1)} > \right) + E^{k+1} - E^k \end{split}$$

where $L = ||f(u)||_{\infty}$ and $E^{k+1} = \frac{1}{2}\epsilon < Au^{(k+1)}, u^{(k+1)} > +\frac{1}{\epsilon} < F(u^{(k+1)} - F(u^{(k)}, 1) > +\frac{1}{\epsilon}$. Finally have, taking the scalar product of (57) with $\sqrt{\Delta t}(u^{(k+1)} - u^{(k)})$,

$$\frac{\|u^{(k+1)} - u^{(k)}\|^2}{\sqrt{\Delta t}} \leq \tau \sqrt{\Delta t}| < B(\mu^{(k+1)} - \mu^{(k)}), u^{(k+1)} - u^{(k)} > | + \sqrt{\Delta t}| < A\mu^{(k)}, u^{(k+1)} - u^{(k)} > |, u^{(k+1)} - u^{(k)}| > |$$

Then, using Young's inequality

$$\begin{split} \frac{\|\boldsymbol{u}^{(k+1)} - \boldsymbol{u}^{(k)}\|^2}{\sqrt{\Delta t}} & \leq \tau \frac{\Delta t}{2} \left(\theta < B(\boldsymbol{\mu}^{(k+1)} - \boldsymbol{\mu}^{(k)}), \boldsymbol{\mu}^{(k+1)} - \boldsymbol{\mu}^{(k)} > + \frac{1}{\theta} < B(\boldsymbol{u}^{(k+1)} - \boldsymbol{u}^{(k)}), \boldsymbol{u}^{(k+1)} - \boldsymbol{u}^{(k)} > \right) \\ & + \frac{\Delta t}{2} \left(\eta < A\boldsymbol{\mu}^{(k)}, \boldsymbol{\mu}^{(k)} > + \frac{1}{\eta} < A(\boldsymbol{u}^{(k+1)} - \boldsymbol{u}^{(k)}), \boldsymbol{u}^{(k+1)} - \boldsymbol{u}^{(k)} > \right) \end{split}$$

We take the sum of these inequalities to get

$$\begin{split} &(\frac{1}{\sqrt{\Delta t}} - \frac{L}{2\epsilon}) \|u^{(k+1)} - u^{(k)}\|^2 \\ &(\epsilon(\tau - \frac{\beta}{2}) - \frac{\tau \Delta t}{2\theta} - \beta \frac{\Delta t}{2\eta}) < B(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} > \\ &+ \frac{\Delta t}{2} (\tau - \beta - \tau \theta) < B(\mu^{(k+1)} - \mu^{(k)}), \mu^{(k+1)} - \mu^{(k)} > \\ &+ \frac{\Delta t}{2} (\tau - \beta - \beta \eta) < B\mu^{(k)}, \mu^{(k)} > \\ &+ \frac{\Delta t}{2} \left(< A\mu^{(k+1)}, \mu^{(k+1)} > + \tau < B\mu^{(k+1)}, \mu^{(k+1)} > \right) + E^{k+1} - E^k \end{split}$$

We can now deduce stability conditions. First of all we assume that

- $\tau(1-\theta) \ge \beta$
- $\tau \ge (1 + \beta)$

Then, the stability is obtained when each factor is positive, say

- $0 < \Delta t < \frac{4\epsilon^2}{L^2}$
- $0 < \Delta t < \frac{\epsilon(\tau \beta/2)}{\frac{\tau}{2\theta} + \frac{\beta}{2\eta}}$

Remark 4.2 Another stability condition can be obtained, inferring that $\langle Bv, v \rangle \leq \rho(B) ||v||^2$ so we use the minoration

$$\begin{split} &\frac{1}{\sqrt{\Delta t}}\|u^{(k+1)}-u^{(k)}\|^2\\ &+(\epsilon(\tau-\frac{\beta}{2})-\rho(B)\frac{L}{2\epsilon})-\frac{\tau\Delta t}{2\theta}-\beta\frac{\Delta t}{2\eta}) < B(u^{(k+1)}-u^{(k)}), u^{(k+1)}-u^{(k)}>\\ &\leq \\ &(\frac{1}{\sqrt{\Delta t}}-\frac{L}{2\epsilon})\|u^{(k+1)}-u^{(k)}\|^2\\ &+(\epsilon(\tau-\frac{\beta}{2})-\frac{\tau\Delta t}{2\theta}-\beta\frac{\Delta t}{2\eta}) < B(u^{(k+1)}-u^{(k)}), u^{(k+1)}-u^{(k)}> \end{split}$$

Finally, we get

•
$$0 < \Delta t < \frac{\epsilon(\tau - \beta/2 - \rho(B)\frac{L}{2\epsilon^2})}{\frac{\tau}{2\theta} + \frac{\beta}{2\eta}}$$

which is a unconditional stability condition satisfied for τ large enough.

We can now give other stability results for nonlinear RSS schemes. We first consider

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \tau B(\mu^{(k+1)} - \mu^{(k)}) + A\mu^{(k)} = 0, \tag{59}$$

$$\mu^{(k+1)} = \epsilon \tau B(u^{(k+1)} - u^{(k)}) + \epsilon A u^{(k)} + \frac{1}{\epsilon} DF(u^{(k+1)}, u^{(k)}). \tag{60}$$

where $DF(u^{(k+1)}, u^{(k)})$ is defined in Section 3.2.

Theorem 4.3 If $\tau > \beta$, then the scheme (59)-(60) is unconditionally stable.

Proof. We proceed exactly as in Theorem 4.1. We obtain after the usual simplifications

$$\begin{split} &\epsilon(\tau - \frac{\beta}{2}) < B(u^{(k+1)} - u^{(k)}), u^{(k+1)} - u^{(k)} > \\ &+ \frac{\Delta t}{2}(\tau - \beta) < B(\mu^{(k+1)} - \mu^{(k)}), \mu^{(k+1)} - \mu^{(k)} > \\ &+ \frac{\Delta t}{2}(\tau - \beta) < B\mu^{(k)}, \mu^{(k)} > + \frac{\Delta t}{2} \left(< A\mu^{(k+1)}, \mu^{(k+1)} > + \tau < B\mu^{(k+1)}, \mu^{(k+1)} > \right) + E^{k+1} - E^k \leq 0, \end{split}$$

where

$$E^{k+1} = \frac{1}{\epsilon} < F(u^{(k+1)}, 1 > +\frac{1}{2}\epsilon < Au^{(k+1)}, u^{(k+1)} >$$

The scheme is then stable when $R^{k+1} < R^k$. Hence the sufficient stability conditions.

Remark 4.4 Stabilization of semi-implicit scheme for Cahn-Hilliard equations have been considered, e.g. in [6, 7, 18] for inpainting problems,

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + A\mu^{(k+1)} + c_1 A(u^{(k+1)} - u^{(k)}) + c_2 (u^{(k+1)} - u^{(k)}) + \lambda_0 D(u^{(k)} - g) = 0, \quad (61)$$

$$\mu^{(k+1)} = \epsilon A u^{(k+1)} + \frac{1}{\epsilon} f(u^{(k)}). \quad (62)$$

Here c_1 and c_2 are positive constants, they play the role of stabilization parameters. Large values of c_1 and c_2 allow to take large time step, however it damages the dynamics. Our approach is here different.

We now describe the practical solution. We can write

$$\begin{pmatrix} Id & \tau \Delta tB \\ -\epsilon \tau B & Id \end{pmatrix} \begin{pmatrix} u^{(k+1)} - u^{(k)} \\ \mu^{(k+1)} \end{pmatrix} = \begin{pmatrix} -\Delta t A u^{(k)} \\ \epsilon A u^{(k)} + \frac{1}{\epsilon} f(u^{(k)}) \end{pmatrix}$$

The matrix of the system can be factorized as Block LU

$$M = \begin{pmatrix} Id & \tau \Delta tB \\ -\epsilon \tau B & Id \end{pmatrix} = \begin{pmatrix} Id & 0 \\ -\epsilon \tau B & Id \end{pmatrix} \begin{pmatrix} Id & \tau \Delta tB \\ 0 & S \end{pmatrix}$$

where $S = Id + \tau^2 \Delta t \epsilon B^2$ is the Schur complement. We have to solve the coupled linear system

$$\begin{cases} X_1 + \tau \Delta t B X_2 = F1, \\ -\tau \epsilon B X_1 + X_2 = F_2. \end{cases}$$

Hence

$$(Id + \tau^2 \Delta t \epsilon B^2) X_2 = F_2 + \epsilon \tau B F 1$$

Then,

$$X_1 = F_1 - \tau \Delta t B X_2$$

We can resume the solution as

Algorithm 4: RSS Cahn-Hilliard

- 1: for $k = 0, 1, \cdots$ until convergence do
- 2: **Set** $F_1 = -\Delta t A \mu^{(k)}$ and $F_2 = -\mu^{(k)} + \epsilon A u^{(k)} + \frac{1}{\epsilon} f(u^{(k)})$
- 3: Solve $(Id + \tau^2 \Delta t \epsilon B^2) \delta \mu = F_2 + \tau \epsilon B F_1$
- 4: **Set** $\mu^{(k+1)} = \mu^{(k)} + \delta \mu$
- 5: Set $u^{(k+1)} = u^{(k)} \tau \Delta t B \delta u$
- 6: end for

When considering the inpainting model, the RSS scheme can be written as

$$\frac{u^{(k+1)} - u^{(k)}}{\Delta t} + \tau B(\mu^{(k+1)} - \mu^{(k)}) + A\mu^{(k)} + \lambda_0 D(u^{(k+1)} - g) = 0, \tag{63}$$

$$\mu^{(k+1)} = \epsilon \tau B(u^{(k+1)} - u^{(k)}) + \epsilon A u^{(k)} + \frac{1}{\epsilon} f(u^{(k)}). \tag{64}$$

say in the matricidal form

$$\begin{pmatrix} Id + \delta t \lambda_0 D & \tau \Delta t B \\ -\epsilon \tau B & Id \end{pmatrix} \begin{pmatrix} u^{(k+1)} - u^{(k)} \\ \mu^{(k+1)} - \mu^{(k)} \end{pmatrix} = \begin{pmatrix} \Delta t (\lambda_0 D(g - u^{(k)}) - Au^{(k)}) \\ \epsilon Au^{(k)} + \frac{1}{\epsilon} f(u^{(k)}) - \mu^{(k)} \end{pmatrix}$$

The implementation of the scheme reads as

```
Algorithm 5: RSS Cahn-Hilliard for inpainting
```

```
1: for k = 0, 1, \cdots until convergence do
2: Set F_1 = \Delta t (\lambda_0 D(g - u^{(k)}) - Au^{(k)})
3: Set F_2 = -\mu^{(k)} + \epsilon Au^{(k)} + \frac{1}{\epsilon} f(u^{(k)})
4: Solve (Id + \lambda_0 D + \tau^2 \Delta t \epsilon B^2) \delta = F_1 - \tau \Delta t B F_2
5: Set \delta \mu = F_2 + \epsilon \tau B \delta
6: Set u^{(k+1)} = u^{(k)} + \delta
7: Set \mu^{(k+1)} = \mu^{(k)} + \delta \mu
8: end for
```

5 Numerical Results

5.1 Implementation

The applications we are interested with are Allen-Cahn and Cahn-Hilliard equations to which homogeneous Neumann boundary conditions are associated. We proceed as in [8] and we first discretize in space the equation with high order finite difference compact schemes; the matrix A corresponds then to the laplacean with Homogeneous Neumann BC (HNBC). Matrix B is the (sparse) second order laplacian matrix with HNBC. For a fast solution of linear systems in the RSS, we will use the cosine-FFT to solve the Neumann problems with matrix $Id + \tau \Delta tB$. test_Neumann_2D.m is a (non RSS) solver that uses cosine-FFT for 2D Neumann problem on the square

5.2 Allen-Cahn equation

Allen_Cahn_fft.m runs (a non RSS) Allen-Cahn with semi-implicit scheme and cosine-FFT, see directory AC CH: this seems correct

AlsoAllen_Cahn_fft_3D.m runs (a non RSS) Allen-Cahn with semi-implicit scheme and cosine-FFT, see directory AC CH: To check

- ALLEN CAHN + SPLITTING + CN-RSS ON THE LINEAR PART RSS splitting AC3D CS.m (WITH FFT)
- ALLEN CAHN 3D + EULER-RSS WITH SPLITTING IN EVERY DIRECTION AC 3D RSS.m (WITHOUT FFT)

APPLICATIONS: PATTERNS, IMAGE RESTAURATION (???) see http://fr.mathworks.com/discove see [22] p 1604.

Method	N	ϵ	Δt	τ	[0,T]	$ error _{\infty}$	CPU factor
RSS	N = 64	0.5	10^{-3}	5	[0, 1]	0.0194	1
RSS	N = 64	0.5	10^{-3}	2	[0, 1]	0.0084	1
Classic	N = 64	0.5	10^{-3}		[0, 1]	0.0047	226
RSS	N = 64	0.5	10^{-2}	2.2	[0, 1]	0.0773	1
Classic	N = 64	0.5	10^{-2}		[0, 1]	0.0486	226

Table 2: 2D Allen-Cahn equation: simulation of patterns - RSS-semi-implicit scheme vs classic semi-implicit scheme, exact solution is $u(x, y, t) = \cos(\pi x) \cos(\pi y) \exp(\sin(3\pi t))$, $\Omega = [0, 1]^2$

5.2.1 Patterns dynamics

Method	N	ϵ	Δt	τ	[0,T]	$ error _{\infty}$	CPU factor
RSS	N = 32	0.5	10^{-3}	5	[0, 1]	5.960^{-2}	1
RSS	N = 32	0.5	10^{-3}	2	[0, 1]	$3.03 \ 10^{-2}$	1
Classic	N = 32	0.5	10^{-3}		[0, 1]	$2.1 \ 10^{-2}$	2.22
RSS	N = 32	0.5	10^{-2}	2	[0, 1]	0.3123	1
RSS	N = 32	0.5	10^{-2}	1.9	[0, 1]	0.3066	1
Classic	N = 32	0.5	10^{-2}		[0, 1]	0.2586	2.22

Table 3: 3D Allen-Cahn equation: simulation of patterns - RSS-Lie splitting scheme vs classic Lie -splitting scheme, exact solution is $u(x,y,z,t) = \cos(\pi x)\cos(\pi y)\cos(\pi z)\exp(\sin(3\pi t))$, $\Omega = [0,1]^3$

Image Segmentation

Image Segmentation consists in labelling pixels in digital images in such a way the image becomes easier to analyse; in particular it allows to locate objects and boundaries. Allen-Cahn equations were proposed as a tool in image segmentation as follows, [22]

$$\frac{\partial \phi}{\partial t} - \Delta \phi + \frac{F'(\phi)}{\epsilon^2} + \lambda \left((1+\phi)(f_0 - c_1)^2 - (1-\Phi)(f_0 - c_2)^2 \right), \quad x \in \Omega$$
 (65)

$$\frac{\partial \Phi}{\partial n} = 0, \quad \partial \Omega \tag{66}$$

If C is the segmenting curve, then the phase ϕ corresponds to the situations

$$\phi(x) = \begin{cases} > 0 & \text{if } x \text{ is inside } C, \\ = 0 & \text{if } x \in C, \\ < 0 & \text{if } x \text{ is outside } C, \end{cases}$$

Here $\epsilon > 0$, $F'(\phi) = \phi(\phi^2 - 1)$, λ is a nonnegative parameter, f_0 is the given image. The terms c_1 and c_2 are the averages of f_0 in the regions $(\phi \geq 0)$ and $(\phi < 0)$, say

$$c_1 = \frac{\int_{\Omega} f_0(x)(1+\phi(x))dx}{\int_{\Omega} (1+\phi(x))dx}$$
 and $c_2 = \frac{\int_{\Omega} f_0(x)(1-\phi(x))dx}{\int_{\Omega} (1-\phi(x))dx}$

These constants will be updated at each time step. In our numerical experiments, the given image f is normalized with $f_0 = \frac{f - f_{min}}{f_{max} - f_{min}}$, where f_{max} and f_{min} are the maximum and the minimum values of the given image, respectively, so we have $f_0 \in [0, 1]$. The initial condition is $\phi = 2f_0 - 1$ and $\Omega = [0, 1]^2$.

We can now write the RSS method

Algorithm 6: RSS splitting for Image segmentation with Allen Cahn

1: **for**
$$k = 0, 1, \dots$$
do

2: Set

$$c_1^{(k)} = \frac{\int_{\Omega} f_0(x)(1 + \phi^{(k)}(x))dx}{\int_{\Omega} (1 + \phi^{(k)}(x))dx},$$

$$c_2^{(k)} = \frac{\int_{\Omega} f_0(x)(1 - \phi^{(k)}(x))dx}{\int_{\Omega} (1 - \phi^{(k)}(x))dx}$$

3: Solve
$$\frac{\phi^{(k+1/3)} - \phi^{(k)}}{\Delta t} = -\lambda \left((1 + \phi^{(k+1/3)})(f_0 - c_1^{(k)})^2 - (1 - \phi^{(k+1/3)})(f_0 - c_2^{(k)})^2 \right)$$

4: Solve
$$(Id + \tau \Delta tB)\delta \phi = -\Delta tA\phi^{(k+/3)}$$

5: **Set**
$$\phi^{(k+2/3)} = \phi^{(k+1/3)} + \delta \phi$$

4: Solve
$$(Id + \tau \Delta tB)\delta \phi = -\Delta tA\phi^{(k+/3)}$$

5: Set $\phi^{(k+2/3)} = \phi^{(k+1/3)} + \delta \phi$
6: Set $\phi^{(k+1)} = \frac{\phi^{(k+2/3)}}{\sqrt{e^{-2\frac{\Delta t}{e^2}} + (\phi^{(k+2/3)})^2(1 - e^{-2\frac{\Delta t}{e^2}})}}$

7: end for

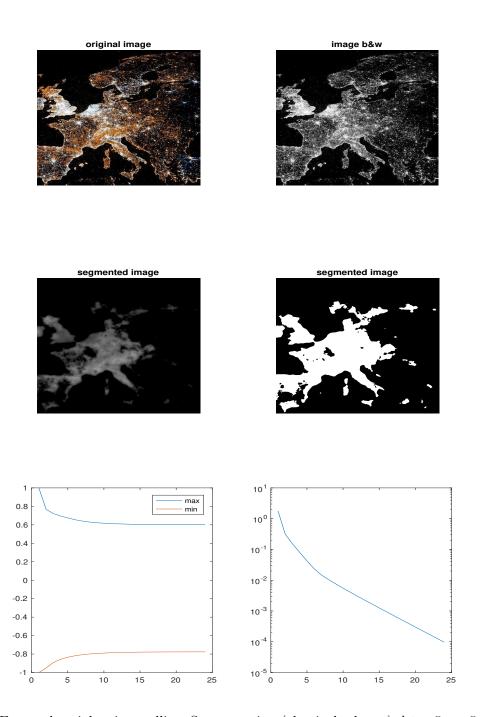


Figure 3: Europe by night via satellite. Segmentation (classical scheme) $\Delta t=5.e-5,\,\epsilon=0.04,\,\tau=1,\,A=B,\,\lambda=750000$

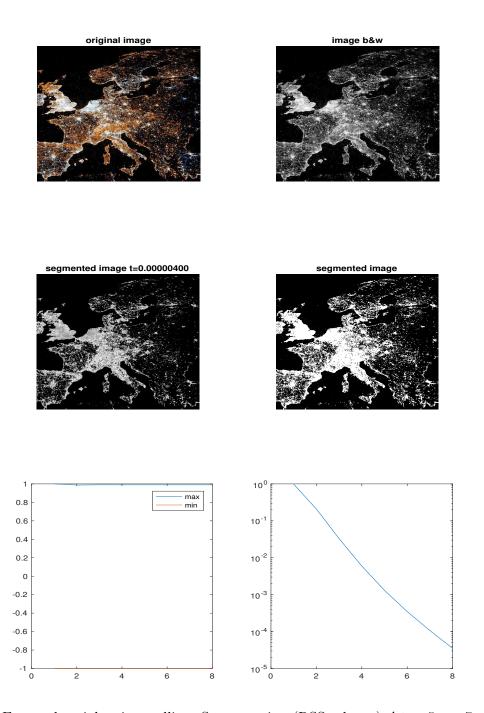


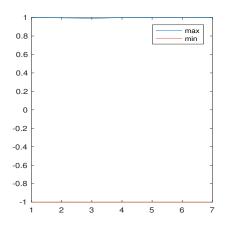
Figure 4: Europe by night via satellite. Segmentation (RSS scheme) $\Delta t=5.e-7,~\epsilon=0.04,~\tau=1,~\lambda=10^{10}$





segmented image





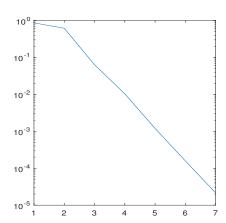


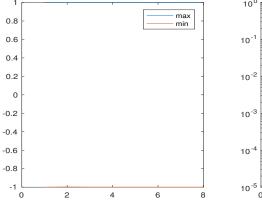
Figure 5: House. Segmentation (RSS scheme) $\Delta t = 1.e-5,\, \epsilon = 0.01,\, \tau = 1,\, \lambda = 10^{10}$





segmented image t=0.00000400





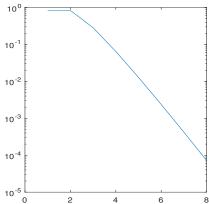


Figure 6: Photograph. Segmentation (RSS scheme) $\Delta t = 5.e - 7, \, \epsilon = 0.04, \, \tau = 1, \, \lambda = 10^{10}$

5.3 Cahn-Hilliard equation

USE THE (NON RSS BUT STABILIZED AS IN BERTOZZI PAPER) CODEs IN THE DIRECTORY CH_INPAINTING

- PATTERNS COMPUTATION: IMPLICIT EXPLIT SCHEME WITH FFT 3DCahn Hilliard fft.m (NON RSS)
- INPAINTING 2D test CH2D RSS solver1.m (without fft), see [9], p22

APPLICATIONS: 2D (3D) PATTERNS, 2D, 3D INPAINTING (SEE WHATEN et al)

5.3.1 Patterns dynamics

Method	N	ϵ	Δt	τ	[0,T]	$\ error\ _{\infty}$	CPU
RSS	N = 64	0.5	10^{-3}	5	[0, 1]	0.0194	7.358853
RSS	N = 64	0.5	10^{-3}	2	[0, 1]	0.0084	7.196025
Classic	N = 64	0.5	10^{-3}		[0, 1]	0.0047	1661.661410
RSS	N = 64	0.5	10^{-2}	2.2	[0, 1]	0.0773	0.795797
Classic	N = 64	0.5	10^{-2}		[0, 1]	0.0486	157.812224

Table 4: 2D Allen-Cahn equation: simulation of patterns - RSS-semi-implicit scheme vs classic semi-implicit scheme, exact solution is $u(x,y,t) = \cos(\pi x)\cos(\pi y)\exp(\sin(3\pi t))$, $\Omega = [0,1]^2$ pgm Allen Cahn fft RSS.m

Method	N	ϵ	Δt	τ	[0,T]	$ error _{\infty}$	CPU
RSS	N = 32	0.5	10^{-3}	5	[0, 1]	5.960^{-2}	94.874912
RSS	N = 32	0.5	10^{-3}	2	[0, 1]	$3.03 \ 10^{-2}$	94.874912
Classic	N = 32	0.5	10^{-3}		[0, 1]	$2.1 \ 10^{-2}$	210.545565
RSS	N = 32	0.5	10^{-2}	2	[0, 1]	0.3123	9.495254
RSS	N = 32	0.5	10^{-2}	1.9	[0, 1]	0.3066	9.819845
Classic	N = 32	0.5	10^{-2}		[0, 1]	0.2586	21.360027

Table 5: 3D Allen-Cahn equation: simulation of patterns - RSS-Lie splitting scheme vs classic Lie -splitting scheme, exact solution is $u(x, y, z, t) = \cos(\pi x) \cos(\pi y) \cos(\pi z) \exp(\sin(3\pi t))$, $\Omega = [0, 1]^3$ pgm Allen Cahn fft.m

5.3.2 2D Inpainting

Method	N	ϵ	Δt	τ	[0,T]	quality	CPU factor (iterations)
RSS	N = 64	0.05	10^{-3}	1.4	[0, 0.1]	EX	1
Classic	N = 64	0.05	10^{-3}		[0, 0.1]	EX	> 10
RSS	N = 64	0.05	5.10^{-3}	1.5	[0, 0.1]	EX	1
Classic	N = 64	0.05	5.10^{-3}		[0, 0.1]	EX	> 10
RSS	N = 64	0.05	10^{-2}	2.8	[0, 0.1]	middle	1
Classic	N = 64	0.5	10^{-2}		[0, 0.1]	middle	> 10

Table 6: 2D Cahn-Hilliard In painting equation, the triangle example: , $\Omega = [0,1]^2,\, \lambda = 90000$

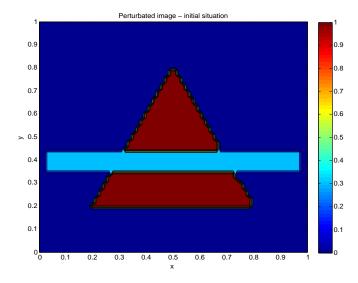


Figure 7: In painting with C-H. $\Delta t = 0.001, \, \epsilon = 0.05, \, N = 64$ - Initial in painted image

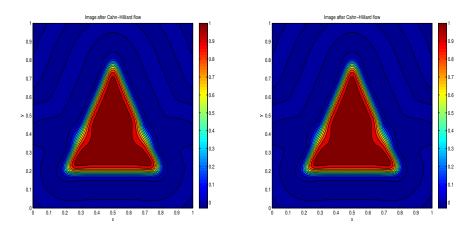


Figure 8: In painting with C-H. $\Delta t=0.001,\,\epsilon=0.05,\,N=64$ - Restored triangle at T=0.1, classical (left) RSS method (right)

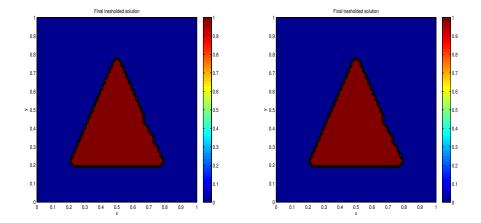


Figure 9: Inpainting with C-H. $\Delta t = 0.001$, $\epsilon = 0.05$, N = 64 - Restored triangle with thresholding at T = 0.1, classical (left) RSS method (right)

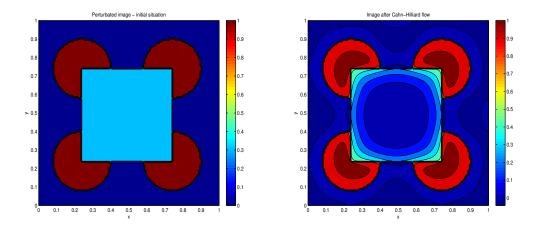


Figure 10: Inpainting with C-H. $\Delta t=0.001,\,\epsilon=0.05,\,N=128$ - Initial inpainted image, left $(t=0),\,\mathrm{right}$ (t=0.005)

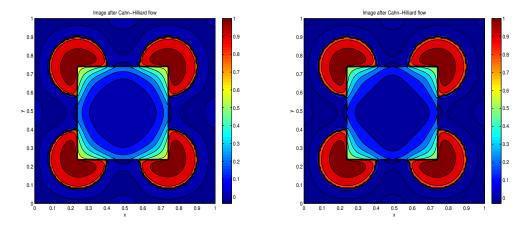


Figure 11: In painting with C-H. $\Delta t=0.001,\,\epsilon=0.05,\,N=128$ - image at t=0.008 (left) and at t=0.01 (right)

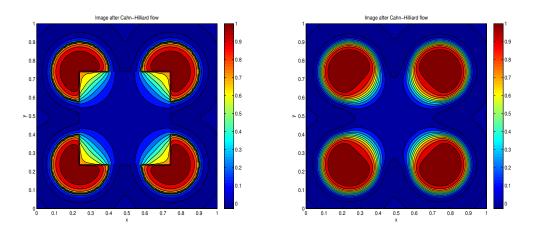


Figure 12: Inpainting with C-H. $\Delta t = 0.001$, $\epsilon = 0.05$, N = 128 - image at t = 0.02 (left) and at t = 0.1 (right). RSS cCheme

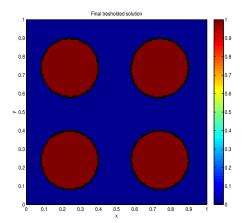


Figure 13: Inpainting with C-H. $\Delta t=0.001,~\epsilon=0.05,~N=128$ - thresholded image at t=0.1.RSSScheme

5.3.3 3D Inpainting

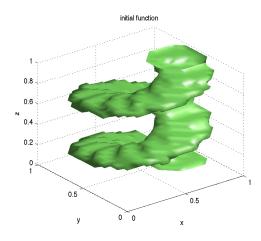


Figure 14: In painting 3D with C-H. $\Delta t = 1.e - 7,\, \lambda = 100000,\, \epsilon = 0.05,\, N = 20$ -

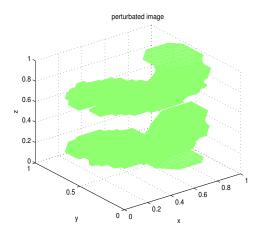


Figure 15: Inpainting 3D with C-H. $\Delta t = 1.e-7,\,\lambda = 100000,\,\epsilon = 0.05,\,N = 20$

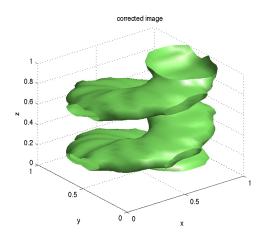


Figure 16: Inpainting 3D with C-H. $\Delta t = 1.e-7,\, \lambda = 100000,\, \epsilon = 0.05,\, N = 20$

6 Concluding Remarks

We have introduced stabilized finite differences semi-implicit schemes that allows to simulate fastly high accurate solution of phase fields problems since the main effort of the computation lies on the efficient solution of sparse linear systems.

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