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## A semi-analytical Fourier spectral method for the Allen-Cahn equation



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#### ABSTRACT

In recent years, Fourier spectral methods have been widely used as a powerful tool for solving phase-field equations. To improve its effectiveness, many researchers have employed stabilized semi-implicit Fourier spectral (SIFS) methods which allow a much larger time step than a usual explicit scheme. Our mathematical analysis and numerical experiments, however, suggest that an effective time step is smaller than a time step specified in the SIFS schemes. In consequence, the SIFS scheme is inaccurate for a considerably large time step and may lead to incorrect morphologies in phase separation processes. In order to remove the time step constraint and guarantee the accuracy in time for a sufficiently large time step, we present a first and a second order semi-analytical Fourier spectral (SAFS) methods for solving the Allen–Cahn equation. The core idea of the methods is to decompose the original equation into linear and nonlinear subequations, which have closed-form solutions in the Fourier and physical spaces, respectively. Both the first and the second order methods are unconditionally stable and numerical experiments demonstrate that our proposed methods are more accurate than the stabilized semi-implicit Fourier spectral method.

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#### 1. Introduction

The phase-field method has recently emerged as a powerful computational approach for modeling and predicting mesoscale morphological and microstructure evolution in materials. This method replaces sharp interfaces by thin but nonzero thickness transition regions where the interfacial forces are smoothly distributed [1]. The basic idea is to introduce conserved or non-conserved order parameters  $\phi$  that vary continuously over thin interfacial layers and are mostly uniform in the bulk phases. The temporal and spatial evolution of the order parameters is governed by the Cahn–Hilliard equation [2–10] or the Allen–Cahn (AC) equation [11–15]. The most significant computational advantage of the phase-field method is that an explicit tracking of the interface is unnecessary.

Numerous numerical algorithms have been developed to improve the accuracy and numerical stability of the phase-field method. There are two types of approaches to achieve high spatial accuracy. One is the adaptive mesh refinement method [16–18] in which additional grid points are placed in regions where the order parameters have large gradients, and the other is the Fourier spectral method [19–30] which uses a spectral representation of a continuous spatial profile of the order parameter. The Fourier spectral method is good for evaluating spatial derivatives since the differential operators are

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represented in the transform space by diagonal matrices that are trivially inverted. Here, we focus on the Fourier spectral method.

To improve the effectiveness of the Fourier spectral method, many researchers have employed stabilized semi-implicit schemes [21,23,24,26,28–30] for the time variable, in which the principal elliptic operator is treated implicitly to reduce the associated stability constraint, while the nonlinear term is still treated explicitly to avoid the expensive process of solving nonlinear equations at each time step [19]. And an extra stabilizing term is added to improve the stability while preserving the simplicity. Stabilized semi-implicit schemes alleviate the time step restriction of explicit schemes, i.e., enable us to use a considerably large time step. But, we observed, through both mathematical analysis and numerical examples, that if we use a large time step, then an effective time step becomes smaller than a time step specified in the stabilized semi-implicit scheme. (Cheng and Rutenberg [31] also pointed out this problem.) In consequence, the stabilized semi-implicit scheme is inaccurate, leading to incorrect morphologies in phase separation processes.

To allow the use of a sufficiently large time step with neither loss of accuracy nor technical difficulty, we present a first and a second order (in time) semi-analytical Fourier spectral methods for solving the AC equation. A core idea of this method is to decompose the original equation into linear and nonlinear subequations, which have closed-form solutions in the Fourier and physical spaces, respectively. In particular, our method requires no additional computational cost and is unconditionally stable in the sense that the solution remains to be bounded for all time regardless of the time step size. It can be used to solve various modified forms of the AC equation such as crystal growth [32,33], grain growth [34–36], and image analysis [37–44].

This paper is organized as follows. In Section 2, we present a first order semi-analytical Fourier spectral method for the AC equation and prove the unconditional stability of this method. We present numerical experiments of the first order stabilized semi-implicit and semi-analytical schemes in Section 3. In Section 4, we introduce the second order stabilized semi-implicit and semi-analytical schemes, and present numerical experiments of both schemes. Conclusions are drawn in Section 5.

#### 2. A semi-analytical Fourier spectral method

We consider the AC equation in two-dimensional space:

$$\frac{\partial \phi(\mathbf{x}, t)}{\partial t} = \frac{\phi(\mathbf{x}, t) - \phi^{3}(\mathbf{x}, t)}{\epsilon^{2}} + \Delta \phi(\mathbf{x}, t), \quad \mathbf{x} \in \Omega, \ 0 < t \le T,$$
(1)

where  $\mathbf{x} = (x_1, x_2)$  is a point inside a rectangular domain  $\Omega = (0, L_1) \times (0, L_2)$ . The boundary condition is the zero Neumann boundary condition:

$$\nabla \phi \cdot \mathbf{n} = 0 \quad \text{on } \partial \Omega, \tag{2}$$

where **n** is the unit normal vector to  $\partial \Omega$ . Instead of using a finite difference approximation, we propose using a Fourier spectral approximation to the AC equation. To solve the AC equation with the boundary condition (2), we employ the discrete cosine transform [45]: for  $k_1 = 0, \ldots, N_1 - 1$  and  $k_2 = 0, \ldots, N_2 - 1$ ,

$$\widehat{\phi}_{k_1 k_2} = \alpha_{k_1} \beta_{k_2} \sum_{l_1=0}^{N_1-1} \sum_{l_2=0}^{N_2-1} \phi_{l_1 l_2} \cos \left[ \frac{\pi}{N_1} k_1 \left( l_1 + \frac{1}{2} \right) \right] \cos \left[ \frac{\pi}{N_2} k_2 \left( l_2 + \frac{1}{2} \right) \right],$$

where  $\alpha_0 = \sqrt{1/N_1}$ ,  $\beta_0 = \sqrt{1/N_2}$ ,  $\alpha_{k_1} = \sqrt{2/N_1}$ ,  $\beta_{k_2} = \sqrt{2/N_2}$  if  $k_1, k_2 \ge 1$ , and  $\phi_{l_1 l_2} = \phi\left(\frac{l_1}{N_1}\left(l_1 + \frac{1}{2}\right), \frac{l_2}{N_2}\left(l_2 + \frac{1}{2}\right)\right)$ . Then, Eq. (1) can be transformed into a set of ordinary differential equations in the cosine space:

$$\frac{\partial \widehat{\phi}_{k_1 k_2}}{\partial t} = \frac{\widehat{\phi}_{k_1 k_2} - \widehat{\phi}^3_{k_1 k_2}}{\epsilon^2} + A_{k_1 k_2} \widehat{\phi}_{k_1 k_2},\tag{3}$$

where  $A_{k_1k_2} = -\pi^2 \left[ \left( \frac{k_1}{L_1} \right)^2 + \left( \frac{k_2}{L_2} \right)^2 \right]$ .

An Explicit Fourier Spectral (ExFS) method is to approximate (3) by

$$\widehat{\phi}_{k_1 k_2}^{n+1} = \widehat{\phi}_{k_1 k_2}^n + \Delta t \left( \frac{\widehat{\phi}_{k_1 k_2}^n - \widehat{(\phi^n)^3}_{k_1 k_2}}{\epsilon^2} + A_{k_1 k_2} \widehat{\phi}_{k_1 k_2}^n \right). \tag{4}$$

This method offers excellent spatial accuracy, but it is only first order accurate in time and suffers from the severe time step restriction for stability. To overcome this problem, many researchers have used the first order stabilized Semi-Implicit Fourier Spectral (SIFS) method [26,28,29]:

$$\left(1 - \Delta t A_{k_1 k_2} + \frac{S \Delta t}{\epsilon^2}\right) \widehat{\phi}_{k_1 k_2}^{n+1} = \widehat{\phi}_{k_1 k_2}^n + \Delta t \frac{(S+1) \widehat{\phi}_{k_1 k_2}^n - \widehat{(\phi^n)^3}_{k_1 k_2}}{\epsilon^2}, \tag{5}$$

where S(>0) is the stabilization parameter. In (5), an extra dissipative term  $\frac{S\Delta t}{\epsilon^2}\left(\widehat{\phi}_{k_1k_2}^{n+1}-\widehat{\phi}_{k_1k_2}^n\right)$  is introduced to improve the stability while preserving the simplicity. The stabilized semi-implicit treatment allows us to use a sufficiently large time step. Eq. (5) implicitly defines the updated phase-field  $\widehat{\phi}_{k_1k_2}^{n+1}$  and can be directly solved in Fourier space to give a Euler-like update

$$\widehat{\phi}_{k_1 k_2}^{n+1} = \widehat{\phi}_{k_1 k_2}^n + \frac{\Delta t}{1 - \Delta t A_{k_1 k_2} + \frac{S \Delta t}{\epsilon^2}} \left( \frac{\widehat{\phi}_{k_1 k_2}^n - \widehat{(\phi^n)^3}_{k_1 k_2}}{\epsilon^2} + A_{k_1 k_2} \widehat{\phi}_{k_1 k_2}^n \right). \tag{6}$$

Eq. (6) can be viewed as an explicit method in the form of (4) with a  $k_1k_2$ -dependent effective time step  $\Delta t_{\text{eff}}$ .

$$\Delta t_{\text{eff}} := \frac{\Delta t}{1 - \Delta t A_{k_1 k_2} + \frac{S \Delta t}{\epsilon^2}} \le \Delta t_{\text{critical}} := \frac{\Delta t}{1 + \frac{S \Delta t}{\epsilon^2}}.$$
 (7)

This implies that  $\Delta t_{\rm eff} \to \Delta t$  as  $\Delta t \to 0$  which provides the consistency and  $\Delta t_{\rm eff} \le \Delta t_{\rm critical} < \frac{\epsilon^2}{5}$  which guarantees the stability of the method. However, the accuracy is significantly lost for a large time step size since  $\Delta t_{\rm eff} \ll \Delta t$  for  $\Delta t \gg \frac{\epsilon^2}{s}$ . We will discuss this in more detail in Section 3.

To remove the time step constraint and guarantee the accuracy in time for a sufficiently large time step, we propose a Semi-Analytical Fourier Spectral (SAFS) method. In this scheme, we decompose (1) into linear and nonlinear subequations on each time step:

$$\frac{\phi_{l_1 l_2}^* - \phi_{l_1 l_2}^n}{\Delta t} = \Delta \phi_{l_1 l_2}^*,\tag{8}$$

$$\frac{\phi_{l_1 l_2}^{n+1} - \phi_{l_1 l_2}^*}{\Delta t} = \frac{\phi_{l_1 l_2}^{n+1} - (\phi_{l_1 l_2}^{n+1})^3}{\epsilon^2},\tag{9}$$

where  $\phi_{l_1 l_2}^*$  is an intermediate solution obtained by solving the linear subequation (8). Eq. (8) can be considered as the first order implicit discretization of the equation  $\phi_t = \Delta \phi$  with an initial condition  $\phi^n$ , which has the following analytic form of the solution in Fourier space:

$$\widehat{\phi}_{k_1 k_2}^* = e^{A_{k_1 k_2} \Delta t} \widehat{\phi}_{k_1 k_2}^n. \tag{10}$$

Next, Eq. (9) can also be considered as the first order implicit discretization of the equation  $\phi_t = \frac{\phi - \phi^3}{\epsilon^2}$  with an initial condition  $\phi^*$ , which can be integrated analytically in the physical space and the exact solution at  $t^{n+1} = t^n + \Delta t$  is given [15,28] as follows:

$$\phi^{n+1}(\mathbf{x}) = \frac{\phi^*(\mathbf{x})}{\sqrt{e^{\frac{-2\Delta t}{\epsilon^2}} + (\phi^*(\mathbf{x}))^2 \left(1 - e^{\frac{-2\Delta t}{\epsilon^2}}\right)}}.$$
(11)

In the SAFS method, the solution  $\phi^{n+1}$  of the AC equation at (n+1)-th time step can be obtained using  $\phi^n$  by the following two sub-steps:

STEP1. Applying the discrete cosine transform to  $\phi_{l_1 l_2}^n$ , we have  $\widehat{\phi}_{k_1 k_2}^*$  by (10). Finally, we return to the physical space by applying the inverse discrete cosine transform,

$$\phi_{l_1 l_2}^* = \mathcal{C}^{-1} \left[ e^{A_{k_1 k_2} \Delta t} \mathcal{C} \left[ \phi_{l_1 l_2}^n \right] \right],$$

where  $\mathcal C$  denotes the discrete cosine transform and  $\mathcal C^{-1}$  its inverse. STEP2. We get the solution  $\phi_{l_1 l_2}^{n+1}$  by the nonlinear time integration (11),

$$\phi_{l_1 l_2}^{n+1} = \frac{\phi_{l_1 l_2}^*}{\sqrt{e^{\frac{-2\Delta t}{\epsilon^2}} + (\phi_{l_1 l_2}^*)^2 \left(1 - e^{\frac{-2\Delta t}{\epsilon^2}}\right)}}$$

with the intermediate solution  $\phi_{l_1 l_2}^*$  obtained in the first step.

It is obvious that the proposed method is unconditionally stable in the sense that the solution remains to be bounded by 1 for all time. In the first step, a function defined as a cosine series

$$\psi^{\Delta t}(x_1, x_2) := \sum_{k_1=0}^{N_1-1} \sum_{k_2=0}^{N_2-1} e^{A_{k_1 k_2} \Delta t} \widehat{\psi^0}_{k_1 k_2} \cos\left(\frac{\pi}{N_1} k_1 x_1\right) \cos\left(\frac{\pi}{N_2} k_2 x_2\right)$$

is a solution of the heat equation with an initial condition  $\psi^0(x_1, x_2)$  thus satisfies the maximum principle,  $\max_{x_1, x_2}$  $|\psi^{\Delta t}(x_1, x_2)| \le \max_{x_1, x_2} |\psi^{\hat{0}}(x_1, x_2)|$ . Therefore,

$$\max_{l_1,l_2} |\phi_{l_1 l_2}^*| = \max_{x_1,x_2} |\phi^*(x_1,x_2)| \le \max_{x_1,x_2} |\phi^n(x_1,x_2)| = \max_{l_1,l_2} |\phi_{l_1 l_2}^n|.$$

In the second step, if we assume that  $|\phi_{l_1l_2}^*| \leq 1$  for each  $l_1, l_2$  then

$$|\phi_{l_1 l_2}^{n+1}| = \frac{|\phi_{l_1 l_2}^*|}{\sqrt{(\phi_{l_1 l_2}^*)^2 + \left(1 - (\phi_{l_1 l_2}^*)^2\right) e^{\frac{-2\Delta t}{\epsilon^2}}}} \leq 1.$$

Therefore, the proposed method is unconditionally stable regardless of the time step size since  $\max_{l_1,l_2} |\phi_{l_1l_2}^{n+1}| \le 1$  for given  $\max_{l_1, l_2} |\phi_{l_1 l_2}^n| \le 1.$ 

#### 3. Numerical experiments

In this section, we present two examples (traveling wave solution in 1D and spinodal decomposition in 2D) in order to numerically demonstrate the stability and convergence of the proposed semi-analytical Fourier spectral (SAFS) method compared to the well-known stabilized semi-implicit Fourier spectral (SIFS) method with S=2.

#### 3.1. Traveling wave solutions

One of the well-known traveling wave solutions of the Allen-Cahn equation is

$$\phi(x,t) = \frac{1}{2} \left( 1 - \tanh \frac{x - 0.5 - st}{2\sqrt{2}\epsilon} \right),\tag{12}$$

where  $s = 3/(\sqrt{2}\epsilon)$  is the speed of the traveling wave. Using this traveling wave solution, we compare the proposed semi-analytical Fourier spectral method with explicit Fourier spectral (ExFS) and stabilized semi-implicit Fourier spectral methods. We set  $\epsilon = 0.03\sqrt{2}$  and compute  $\phi(x,t)$  for  $0 < t \le T_f = 1/s$  with an initial condition  $\phi(x,0)$  on the domain  $\Omega = [0,4]$ . The numerical solutions are obtained by the ExFS, SIFS, and SAFS methods with various time step sizes  $\Delta t$  but the spatial grid size is fixed to  $h = 2^{-5}$  which provides enough spatial accuracy.

Fig. 1 shows a table for the relative discrete  $l_2$  errors of the numerical solutions  $\phi(x, T_f)$  for different time step sizes and two plots for the computed results with  $\Delta t/T_f=2^{-7}$  and  $2^{-3}$ . All three methods show the first order convergence when  $\Delta t$  is small enough, say  $\Delta t/T_f<2^{-7}$ . The ExFS method, however, has severe stability constraint and  $\Delta t/T_f=2^{-7}$  is close to the maximum time step to maintain numerical stability for  $\epsilon=0.03\sqrt{2}$  while  $\Delta t/T_f=2^{-3}$  is still a stable choice for both the SIFS and the SAFS methods. When  $\Delta t$  is getting bigger, say  $\Delta t/T_f=2^{-7}$  or  $2^{-3}$ , the simulations indicate that the results obtained from the SIFS method travel with a wrong speed unlike the SAFS method, which phenomenon needs to be further investigated.

To clarify the effect of slow moving speed of the SIFS solution, we check an interfacial x position which corresponds to the 0.5-level of numerical solution for different time steps. Let  $x^I(T_f)$  be an interfacial point where  $\phi(x^I, T_f) = 0.5$  and the final time  $T_f = 1/s$  was chosen to make the x position of the exact solution to be 1.5. Fig. 2(a) shows the absolute values of the differences between the exact  $x^{l}$  position (=1.5) and the numerical  $x^{l}$  position with the SIFS and SAFS methods for different time steps. Both methods provide the first order convergence as expected when  $\Delta t$  is small; however, the SIFS method has significant error when  $\Delta t/T_f > 2^{-5}$  or roughly  $\Delta t > 10^{-3}$ . As the time step increases, the difference between the x position of the SIFS method and the exact position 1.5 becomes bigger and bigger.

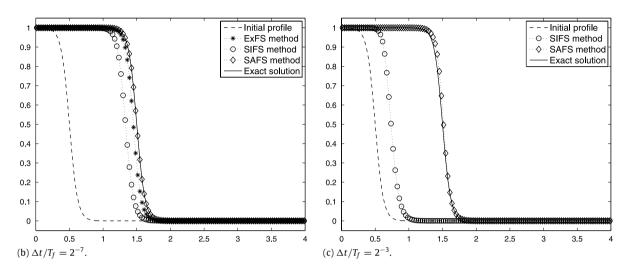
Fig. 2(b) shows the numbers of time steps for a numerical solution to reach the interfacial position  $x^l = 1.5$  as a function of time steps size. The number of time steps for the SAFS method is  $T_f/\Delta t$  as expected; however, the number of time steps for the SIFS method is not approaching to 1 as  $\Delta t \to T_f$ . This means that the effective time step size of the SIFS method for a large time step size is much smaller than the time step size used. The numerical simulation confirms that the number of required time steps can be well estimated by  $\Delta t_{\text{critical}}$  defined in (7),

$$N_{\text{steps}} \approx \frac{T_f}{\Delta t_{\text{eff}}} \geq \frac{T_f}{\Delta t_{\text{critical}}} = \frac{T_f}{\Delta t} + T_f \frac{S}{\epsilon^2},$$

where  $T_f \frac{S}{\epsilon^2}$  in this simulation is  $\frac{\sqrt{2}\epsilon}{3} \frac{2}{\epsilon^2} = \frac{200}{9}$ . Fig. 3 shows the number of time steps to make the numerical solutions reach the correct interfacial position. In order to see the number of time steps only as a function of interface thickness parameter  $\epsilon$ , we choose  $T_f = 1/s = \sqrt{2}\epsilon/3$ ,  $x^I(T_f)=1.5$ , and  $\Delta t=T_f/10$  or  $\Delta t=T_f/1$ . The numerical simulation clearly shows that  $N_{\text{steps}}$  of the SIFS method can be well estimated by  $\frac{T_f}{\Delta t_{\text{critical}}}=\frac{T_f}{\Delta t}+\frac{\sqrt{2}}{3}\frac{S}{\epsilon}$  while  $N_{\text{steps}}$  of the SAFS method is simply  $\frac{T_f}{\Delta t}$ .

$\Delta t/T_f$	$2^{-11}$	$2^{-9}$	$2^{-7}$	$2^{-5}$	$2^{-3}$
ExFS method	4.6729e–3	1.8476e–2	7.0354e-2	unstable	unstable
SIFS method	1.7044e-2	6.5457e-2	2.1761e-1	4.7416e-1	6.6772e-1
SAFS method	1.2188e-4	4.8671e-4	1.9334e-3	7.5161e <del>-</del> 3	2.6270e-2

(a) Relative  $l_2$  errors of  $\phi(x, T_f)$  for various time step sizes.



**Fig. 1.** Computational results of the traveling wave solution  $\phi(x, T_f)$  at  $T_f = 1/s$  with  $\epsilon = 0.03\sqrt{2}$  and  $h = 2^{-5}$ . (a) Relative  $l_2$  errors for the ExFS, the SIFS, and the SAFS methods for various time step sizes. (b) Comparison of the results for  $\Delta t/T_f = 2^{-7}$ . (c) Comparison of the results for  $\Delta t/T_f = 2^{-3}$ .

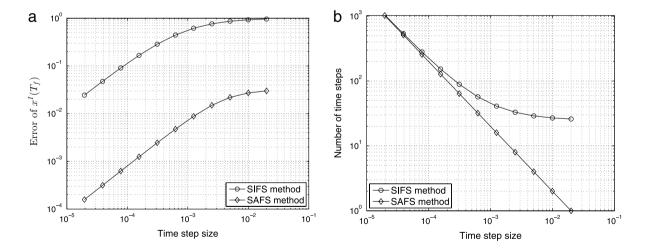
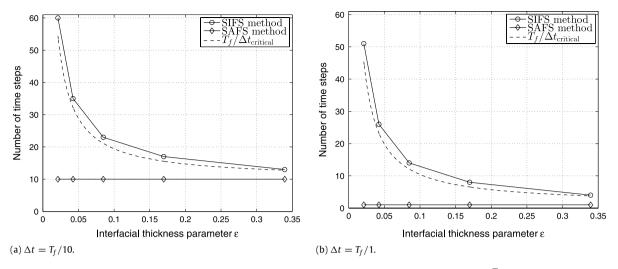


Fig. 2. Interfacial position  $x^l(T_f)$  of numerical solutions at  $T_f = 1/s$  with  $\epsilon = 0.03\sqrt{2}$  for various time step sizes. (a) Errors of the numerical  $x^l(T_f)$  position. (b) Numbers of time steps required for a numerical solution to reach  $x^l = 1.5$ .

#### 3.2. Spinodal decomposition

The Allen-Cahn equation also arises [46] from minimization of the Ginzburg-Landau free energy

$$\mathcal{E}(\phi) := \int_{\varOmega} \left( \frac{F(\phi)}{\epsilon^2} + \frac{|\nabla \phi|^2}{2} \right) d\mathbf{x},$$



**Fig. 3.** Numbers of required time steps for the first order SIFS solution (S=2) to reach  $x^I=1.5$  with  $T_f=1/s=\sqrt{2}\epsilon/3$  as a function of thickness parameter  $\epsilon$ .

where  $F(\phi) = 0.25(1 - \phi^2)^2$  is the bulk free energy having a double well form. The AC equation is the  $L^2$ -gradient flow of the total free energy  $\mathcal{E}(\phi)$ :

$$\frac{\partial \phi}{\partial t} = -\operatorname{grad}\mathcal{E}(\phi),\tag{13}$$

where the symbol "grad" denotes the gradient on the manifold in  $L^2(\Omega)$  space. Let the domain of definition for the functional  $\mathcal{E}$  be  $\mathcal{D}=\{\phi\in H^2(\Omega)\mid \frac{\partial\phi}{\partial\mathbf{n}}=0 \text{ on }\partial\Omega\}$ . Let  $\phi,\psi\in\mathcal{D}$ . Then, we have

$$\begin{split} (\mathrm{grad} \mathcal{E}(\phi), \, \psi)_{L^2} &= \left. \frac{d}{d\theta} \mathcal{E}(\phi + \theta \, \psi) \right|_{\theta = 0} = \lim_{\theta \to 0} \frac{1}{\theta} (\mathcal{E}(\phi + \theta \, \psi) - \mathcal{E}(\phi)) \\ &= \int_{\Omega} \left( \frac{F'(\phi)}{\epsilon^2} - \Delta \phi \right) \psi \, d\mathbf{x} = \left( \frac{F'(\phi)}{\epsilon^2} - \Delta \phi, \, \psi \right)_{L^2}, \end{split}$$

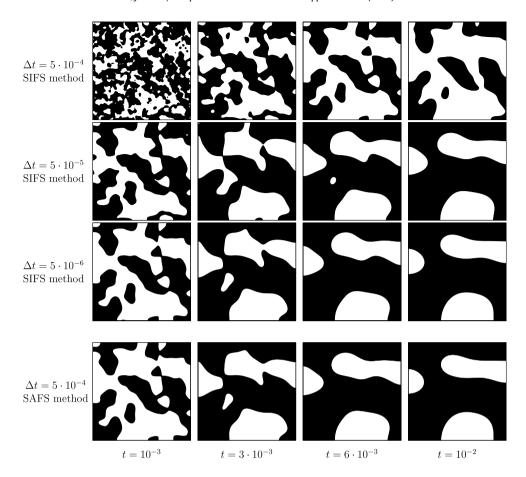
where we have used integration by parts and the boundary condition (2). We identify  $\operatorname{grad}\mathcal{E}(\phi) \equiv F'(\phi)/\epsilon^2 - \Delta\phi$ ; then (13) becomes the AC equation.

For  $\phi_1 = -1/\sqrt{3}$  and  $\phi_2 = 1/\sqrt{3}$ ,  $F''(\phi) > 0$  for  $-1 < \phi < \phi_1$  or  $\phi_2 < \phi < 1$  and  $F''(\phi) < 0$  for  $\phi_1 < \phi < \phi_2$ . The intervals  $(-1,\phi_1)$  and  $(\phi_2,1)$  are called metastable intervals and  $(\phi_1,\phi_2)$  is called the spinodal interval [47]. It is known that  $\phi$  which lies in the spinodal interval is very unstable and that the growth of instabilities results in phase separation, which is called spinodal decomposition. In this subsection, we perform a spinodal decomposition computation to compare the SAFS method with the SIFS method for different time steps. We set  $\epsilon = 0.015$ , h = 1/128 on the domain  $\Omega = [0,1] \times [0,1]$  and the initial condition is  $\phi(x,y,0) = 0.005 \cdot \text{rand}(x,y)$  where rand(x,y) is a random number between -1 and 1.

Fig. 4 shows evolutions of the numerical solution  $\phi(x,y,t)$  using the SIFS and SAFS methods for different time steps. Rows 1, 2, and 3 show snapshots of the solution obtained from the SIFS method at different times for  $\Delta t = 5 \cdot 10^{-4}$ ,  $5 \cdot 10^{-5}$ , and  $5 \cdot 10^{-6}$ , respectively. Row 4 shows snapshots of the solution obtained from the SAFS method at different times for  $\Delta t = 5 \cdot 10^{-4}$ . The results in Fig. 4 show that the SIFS method is robust but significantly inaccurate for a larger time step size, leading to incorrect morphologies in the phase separation process, while there are nearly no significant differences of the SAFS solutions with  $\Delta t \leq 5 \cdot 10^{-4}$  compared to the results plotted with the coarsest time step. It is worthwhile to note that the SIFS method experiences a slow evolution side effect as  $\Delta t$  is getting bigger and the solution computed with  $\Delta t = 5 \cdot 10^{-4}$  at  $t = 6 \cdot 10^{-3}$  is quite similar with the converged solution at  $t = 10^{-3}$ . In fact, the slow-down factor defined as

$$\frac{\Delta t}{\Delta t_{\rm eff}} \approx \frac{\Delta t}{\Delta t_{\rm critical}} = 1 + S \frac{\Delta t}{\epsilon^2} \tag{14}$$

is  $5\frac{4}{9}\approx 5.444$  for the computation with  $\Delta t=5\cdot 10^{-4}$  and  $\epsilon=0.015$ . This simulation numerically demonstrates that the SAFS method has no slow-down side effect and is significantly more accurate than the widely used SIFS method for solving phase-field equations.



**Fig. 4.** Spinodal decomposition of the numerical solution  $\phi(x, y, t)$  obtained from the SIFS and SAFS methods. Each row of the figure shows snapshots of the solution at the times indicated at the bottom and is calculated using the time step indicated on the left hand side of the figure.

#### 4. Second order Fourier spectral methods

One of the important merits of using the Semi-Analytical Fourier Spectral (SAFS) method is the flexibility to develop temporally high order methods by combining operator splitting steps. We present a second order SAFS method which combines a half-time linear solver and a full-time nonlinear solver followed by a final half-time linear solver:

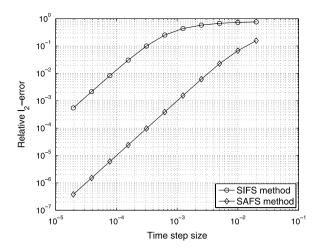
$$\phi^{n} = \phi(\mathbf{x}, t^{n}) \xrightarrow{\phi_{t} = \Delta\phi} \phi^{*} := \phi\left(\mathbf{x}, t^{n} + \frac{\Delta t}{2}\right),\tag{15}$$

$$\phi^* = \phi(\mathbf{x}, t^n) \xrightarrow{\phi_t = (\phi - \phi^3)/\epsilon^2} \phi^{**} := \phi(\mathbf{x}, t^n + \Delta t), \tag{16}$$

$$\phi^{**} = \phi\left(\mathbf{x}, t^n + \frac{\Delta t}{2}\right) \xrightarrow{\phi_t = \Delta \phi} \phi^{n+1} := \phi(\mathbf{x}, t^n + \Delta t), \tag{17}$$

where  $\phi^*$  and  $\phi^{**}$  are the intermediate solutions by solving the subequations on the top of the arrows with the initial conditions on the left of the arrows in (15) and (16), respectively. Eqs. (15) and (17) are similar to (8) and can be solved in a similar way. Also (16) is similar to (9) and can be solved in a similar way. This method is merely an implementation of the operator splitting schemes and provides a second order convergence [28] since each of the sub-steps is computed semi-analytically. The second order method is unconditionally stable and the proof of the stability is trivial if we follow the argument at the end of Section 2.

In this section, we present three numerical examples (again the traveling wave solution in 1D, motion by mean curvature in 2D, and spinodal decomposition 3D) in order to demonstrate the numerical convergence and stability of the method. For the comparison purpose, we also provide numerical solutions by the well-known second-order stabilized semi-implicit Fourier spectral (SIFS) method defined as follows,



**Fig. 5.** Relative  $l_2$  errors of the traveling wave solution at  $T_f = 1/s$  by the SIFS and the SAFS methods for various time step sizes with  $h = 2^{-5}$  and  $\epsilon = 0.03\sqrt{2}$ .

$$\left(\frac{3}{2} - \Delta t A_{k_1 k_2} + \frac{S \Delta t}{\epsilon^2}\right) \widehat{\phi}_{k_1 k_2}^{n+1} = 2 \widehat{\phi}_{k_1 k_2}^n + 2 \Delta t \frac{(S+1) \widehat{\phi}_{k_1 k_2}^n - \widehat{(\phi^n)^3}_{k_1 k_2}}{\epsilon^2} - \frac{1}{2} \widehat{\phi}_{k_1 k_2}^{n-1} - \Delta t \frac{(S+1) \widehat{\phi}_{k_1 k_2}^{n-1} - \widehat{(\phi^{n-1})^3}_{k_1 k_2}}{\epsilon^2}.$$
(18)

#### 4.1. Traveling wave solutions

We solve the AC equation by the second order SAFS method defined in (15)–(17) to obtain the traveling wave solutions  $\phi(x,t)=\frac{1}{2}\left(1-\tanh\frac{x-0.5-st}{2\sqrt{2}\epsilon}\right)$  with the speed of the traveling wave  $s=3/(\sqrt{2}\epsilon)$ . For the comparison purpose, we use the second order SIFS method defined in (18) with S=2. Fig. 5 shows the computational error of the numerical solutions at  $T_f=1/s$  by the SIFS and SAFS methods. Parameters used in this example are the same as those in Section 3.1,  $h=2^{-5}$  and  $\epsilon=0.03\sqrt{2}$ . Both of them provide the second order of accuracy; however, the SAFS starts to converge for a much bigger time step size and gives 2–3 more digits of accuracy where both methods in the convergence region.

#### 4.2. Motion of a circle by its mean curvature

One of the well analyzed solutions of the AC equation is the motion of a circle. Suppose a radially symmetric initial condition is given as follows on the domain  $\Omega = [0, 1] \times [0, 1]$ ,

$$\phi(x, y, 0) = \tanh \frac{0.25 - \sqrt{(x - 0.5)^2 + (y - 0.5)^2}}{\sqrt{2}\epsilon},$$
(19)

which represents a circle centered at (0.5, 0.5) with a radius  $R_0 = 0.25$ . It is well known that the solution of the AC equation with the initial condition is radially symmetric and the radius of the interfacial circle (where  $\phi(x, y, t) = 0$ ) is shrinking by the rate of the curvature of the circle.

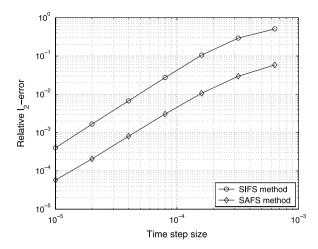
We compute this two-dimensional problem with  $\epsilon=0.01$  using the second order SIFS (with S=2) and SAFS methods on a regular  $256\times256$  spatial grid. Fig. 6 shows the convergence rate of the solutions at  $T_f=0.0256$  by the methods. Here the error is computed by comparison with a quadruplely over-resolved numerical reference solution. Both of the methods converge with second order; however the SAFS starts to converge for a much bigger time step size and gives order of magnitude higher accuracy where both methods in the convergence region.

Asymptotic analysis shows that, in the limit that the radius of the circle is much larger than the interfacial thickness, the interfacial velocity V(t) is given by

$$V(t) = \frac{dR(t)}{dt} = -\frac{d-1}{R(t)}$$

and the radius of the circle at time t can be computed as

$$R(t) = \sqrt{R_0^2 - 2(d-1)t} \quad \text{for } t \in [0, R_0^2/(2(d-1))], \tag{20}$$



**Fig. 6.** Relative  $l_2$  errors of the motion by curvature at  $T_f = 0.0256$  by the SIFS and the SAFS methods for various time step sizes with  $h = 2^{-8}$  and  $\epsilon = 0.01$ 

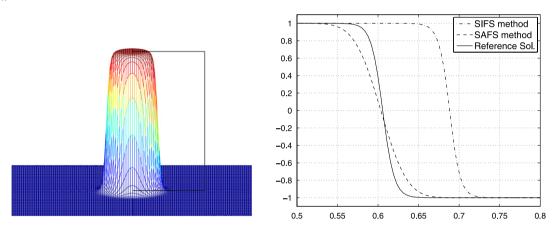


Fig. 7. Reference solution at  $T_f=0.0256$  and cross-section of the solution near the interface by the SIFS and the SAFS methods with  $\Delta t=6.4\cdot 10^{-4}$ .

where  $R_0$  is the initial radius of the circle and d is the number of dimensions. The circle will shrink to the center of the circle and disappear for times larger than  $R_0^2/(2(d-1))$ .

Fig. 7 shows the reference solution at  $T_f = 0.0256$  and two numerical solutions by the SIFS and the SAFS methods using a coarse time step  $\Delta t = 6.4 \cdot 10^{-4}$ . The interfacial circle of the SIFS solution slowly evolves in time and the computed radius R = 0.1884 at  $T_f = 0.0256$  corresponds to the analytic solution at t = 0.0135. The radii of a circle as a function of time obtained from the SIFS and SAFS methods (using  $\Delta t = 1.6 \cdot 10^{-4}$ ,  $3.2 \cdot 10^{-4}$ ,  $6.4 \cdot 10^{-4}$ ) are shown in Fig. 8(a) and (b), respectively. The SIFS method is quite accurate if a smaller time step is employed; however, at a larger time step, the SIFS method leads to a bad accuracy as compared to the analytical result. On the other hand, there is no visible difference between the numerical results obtained from the SAFS method and the analytical result even at larger time steps. This example numerically demonstrates that our SAFS method provides a second order of convergence and does not experience the time-delay side effect unlike the SIFS method.

#### 4.3. Spinodal decomposition in 3-dimension

In this subsection, we perform a spinodal decomposition computation in 3-dimensional space using the second order SAFS method. The governing equation is again the AC equation derived in Section 3.2. We set  $\epsilon = 0.015$  and h = 1/64 on the domain  $\Omega = [0, 1] \times [0, 1] \times [0, 1]$  and the initial condition is  $\phi(x, y, z, 0) = 0.005 \cdot \text{rand}(x, y, z)$  where rand(x, y, z) is a random number between -1 and 1. We have checked that the method has the second order of convergence with respect to the time step size  $\Delta t$  and  $\Delta t = 5 \cdot 10^{-4}$  for the computation is rather large but gives around a couple of digits of accuracy. Fig. 9 shows the computational results at t = 0,  $10^{-3}$ ,  $3 \cdot 10^{-3}$ ,  $6 \cdot 10^{-3}$ ,  $10^{-2}$ ,  $3 \cdot 10^{-2}$  with  $\Delta t = 5 \cdot 10^{-4}$ .

#### 5. Conclusions

The stabilized semi-implicit Fourier spectral method is employed by many researchers since the stabilized semi-implicit scheme allows a much larger time step than a usual explicit scheme. But, we found that the first order stabilized semi-

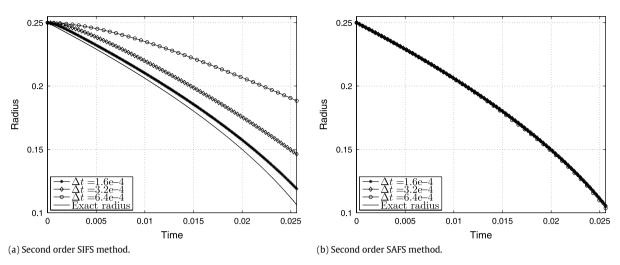
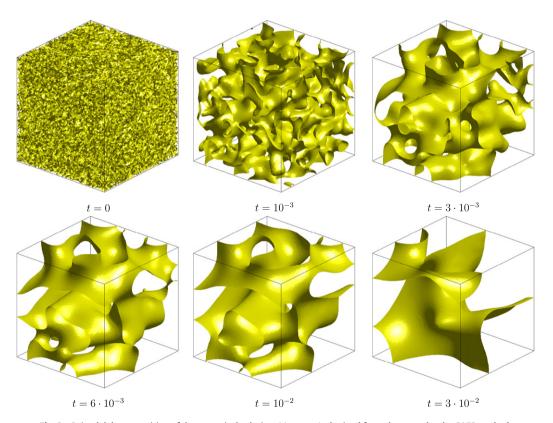


Fig. 8. Radii of the interfacial circle as a function of time  $(0 \le t \le 0.0256)$  obtained by the SIFS and SAFS methods using  $\Delta t = 1.6 \cdot 10^{-4}, 3.2 \cdot 10^{-4}, 6.4 \cdot 10^{-4}$ .



**Fig. 9.** Spinodal decomposition of the numerical solution  $\phi(x, y, z, t)$  obtained from the second order SAFS method.

implicit scheme has a  $k_1k_2$ -dependent effective time step  $\Delta t_{\rm eff}$  as  $\Delta t_{\rm eff} = \frac{\Delta t}{1-\Delta tA_{k_1k_2}+\frac{S\Delta t}{\varepsilon^2}}$ , where  $\Delta t$  is an algorithmic time step. This implies that the effective time step is identical to the algorithmic time step for small enough  $\Delta t$ . However, for a sufficiently large time step, the accuracy is significantly lost since  $\Delta t_{\rm eff} \ll \Delta t$  for  $\Delta t \gg \frac{\epsilon^2}{5}$ . In order to remove the time step constraint and guarantee the accuracy in time for a sufficiently large time step, we presented a semi-analytical Fourier spectral method for solving the Allen–Cahn equation. The core idea of this method was to decompose the original equation into linear and nonlinear subequations, which have closed-form solutions in the Fourier and physical spaces, respectively. This first order method is unconditionally stable and we demonstrated, through numerical experiments, that our proposed first order method is more accurate than the first order stabilized semi-implicit scheme. We also suggested a second order semi-analytical Fourier spectral method which is also unconditionally stable and more accurate than the well-known second

order stabilized semi-implicit method. We expect that our proposed methods can be used to solve various modified forms of the Allen–Cahn equation such as crystal growth, grain growth, and image analysis.

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