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Method 1 for time stepping

This adaptive time-stepping method considers the time derivative of the total energy for details: see Zhang, Z., & Qiao, Z. (2012). An Adaptive Time-Stepping Strategy for the Cahn-Hilliard Equation. Communications in Computational Physics, 11(4), 1261-1278. doi:10.4208/cicp.300810.140411s

$$\Delta t = \max(\Delta t_{\min}, \frac{\Delta t_{\max}}{\sqrt{1+\alpha|e'(t)|^2}})$$

Import

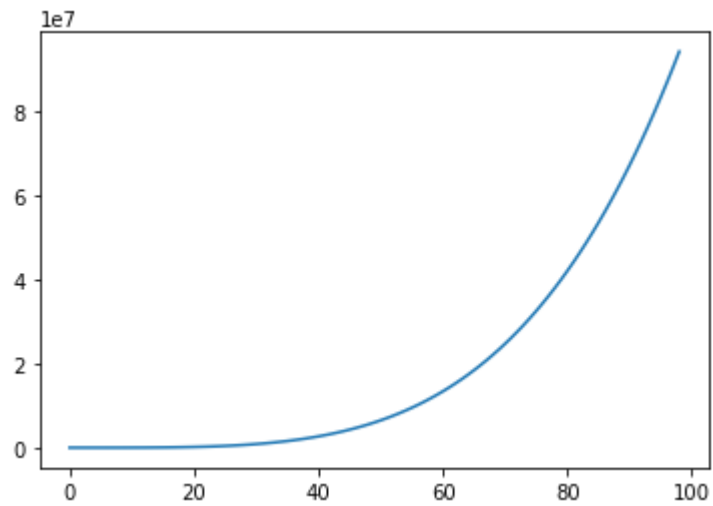
```
In [1]: 1 import numpy as np
        2 import matplotlib.pyplot as plt
        3 import time
        4 import warnings
        5 warnings.simplefilter("ignore", np.ComplexWarning)
        6 import math
        7 import pyvista as pv
        8 import pyfftw          # use for fast fourier transform
        9 from scipy.fft import fft, ifft
       10 from numba import jit  # use to speed up
       11 import scipy.stats as st
```

Functions

In [5]:

```
1 A=1
2 x=np.arange(1,100)
3 f=A*x**2*(1-x)**2
4
5 plt.plot(f)
```

Out[5]: [<matplotlib.lines.Line2D at 0x25695b09dc0>]



In [2]:

```

1 def free_energ(c):
2     A=1.0
3     dfdc =A*(2.0*c*(1-c)**2 -2.0*c**2 *(1.0-c))
4     return dfdc
5
6
7 #-----
8
9 def fft_(a):
10     """
11     return a fft object from pyfftw library that will be use to compute fft
12     """
13     fft_object=pyfftw.builders.fftn(a,axes=(0,1,2), threads=12)
14     return fft_object()
15 #-----
16
17 def ifft_(a):
18     """
19     return a inverse fft object from pyfftw library that will be use to compute inverse fft
20     """
21     ifft_object=pyfftw.builders.ifftn(a,axes=(0,1,2), threads=12)
22     return ifft_object()
23 #-----
24 #@jit(nopython=True)
25 def micro_ch_pre(Nx,Ny,Nz,c0):
26     c=np.zeros((Nx,Ny,Nz))
27     noise=0.02
28     for i_x in range(Nx):
29         for i_y in range(Ny):
30             for i_z in range(Nz):
31                 c[i_x,i_y,i_z] =c0 + noise*(0.5-np.random.rand())
32     return c
33
34 # Compute energy evolution
35 """
36 def calculate_energ(Nx,Ny,Nz,c,grad_coef):
37     energ =0.0
38     for i in range (Nx-1):
39         ip = i + 1
40         for j in range (Ny-1):
41             jp = j + 1
42             if (Nz>1):

```

```

43         for l in range (Nz-1):
44             lp = l + 1
45             energ += c[i,j,l]**2 *(1.0-c[i,j,l])**2 + 0.5*grad_coef*((c[ip,j,l]-c[i,j,l])**2 \
46                 +(c[i,jp,l]-c[i,j,l])**2 + (c[i,j,lp]-c[i,j,l])**2)
47         else:
48             energ += c[i,j,0]**2 *(1.0-c[i,j,0])**2 + 0.5*grad_coef*((c[ip,j,0]-c[i,j,0])**2 \
49                 +(c[i,jp,0]-c[i,j,0])**2)
50
51     return energ
52 """
53 # Compute energy evolution
54 def calculate_energy(Nx,Ny,Nz,c,grad_coef):
55     energ =0.0
56     # -----Nx-----
57     for i in range (Nx-1):
58         ip = i + 1
59         #-----Ny-----
60         if (Ny>1):
61             for j in range (Ny-1):
62                 jp = j + 1
63                 # -----Nz-----
64                 if (Nz>1): # 3D
65                     for l in range (Nz-1):
66                         lp = l + 1
67                         energ += c[i,j,l]**2 *(1.0-c[i,j,l])**2 + 0.5*grad_coef*((c[ip,j,l]-c[i,j,l])**2 \
68                             +(c[i,jp,l]-c[i,j,l])**2 + (c[i,j,lp]-c[i,j,l])**2)
69
70                     else: # (Nz==1) 2D
71                         energ += c[i,j,0]**2 *(1.0-c[i,j,0])**2 + 0.5*grad_coef*((c[ip,j,0]-c[i,j,0])**2 \
72                             +(c[i,jp,0]-c[i,j,0])**2)
73                 else : # (Ny==1):
74                     # -----Nz-----
75                     if (Nz>1): # 2D
76                         for l in range (Nz-1):
77                             lp = l + 1
78                             energ += c[i,0,l]**2 *(1.0-c[i,0,l])**2 + 0.5*grad_coef*((c[ip,0,l]-c[i,0,l])**2 \
79                                 + (c[i,0,lp]-c[i,0,l])**2)
80                     else : # (Nz==1) 1D
81                         energ += c[i,0,0]**2 *(1.0-c[i,0,0])**2 + 0.5*grad_coef*(c[ip,0,0]-c[i,0,0])**2
82
83     return energ
84
85     #-----

```

```

86 def infinite_norm(M):
87     # suppose that Nz=1
88     m=M.shape[0]
89     n=M.shape[1]
90     array_row_sum=[]
91     for i in range(m):
92         row_sum=0
93         for j in range(n):
94             row_sum+=np.abs(M[i,j,0])
95         array_row_sum.append(row_sum)
96     inf_norm=np.max(array_row_sum)
97     return inf_norm
98 # compute 2D laplacian
99 def laplacian_2D(c,i,j,l,dx,dy,dz):
100     ip=i+1
101     im=i-1
102     jp=j+1
103     jm=j-1
104     if (im==-1):
105         im=Nx-1
106     if (ip==Nx):
107         ip=0
108     if (jm==-1):
109         jm=Ny-1
110     if (jp==Ny):
111         jp=0
112     hne=c[ip,j,0]
113     hnw=c[im,j,0]
114     hns=c[i,jm,0]
115     hnn=c[i,jp,0]
116     hnc=c[i,j,0]
117     laplacian=(hnw+hne+hns+hnn-4*hnc)/(dx*dy) # uniform mesh
118     return laplacian
119
120 # compute gradient
121 def gradient(c,i,j,l,dx,dy,dz):
122     Nx=c.shape[0]
123     Ny=c.shape[1]
124     Nz=c.shape[2]
125     im=i-1
126     jm=j-1
127     lm=l-1
128     if (im==-1):

```

```

129     im=Nx-1
130     if (jm==-1):
131         jm=Ny-1
132     if (lm==-1):
133         lm=Nz-1
134
135     hw=c[im,j,l]
136     hs=c[i,jm,l]
137     hz=c[i,j,lm]
138     hc=c[i,j,l]
139
140     grad_mu_x=(hc-hw)/dx
141     grad_mu_y=(hc-hs)/dx
142     grad_mu_z=(hc-hz)/dx
143
144     return grad_mu_x,grad_mu_y,grad_mu_z
145
146     #-----
147     # compute residual at each time step (if istep >1), using Fourier space
148     def compute_residual_fft(array_energy,dtime, dfdck,ck,istep,k,k2, kappa,Nx,Ny,Nz):
149         Nx=c.shape[0]
150         Ny=c.shape[1]
151         Nz=c.shape[2]
152
153         # compute \mu
154         mu=dfdck+kappa*(k2)*ck
155
156
157         grad_mu=(1j*k)*mu
158
159         sum_L2_square=0
160         # compute Nabla(\mu) at each point grid and the associated L2 norm **2
161         for i in range(Nx):
162             for j in range(Ny):
163                 for l in range(Nz):
164                     grad_mu=np.array([ (1j*k[0,i,j,l])*mu[i,j,l] , (1j*k[1,i,j,l])*mu[i,j,l] , (1j*k[2,i,j,l])*
165                     grad_mu=np.real(np.fft.ifft(grad_mu))
166                     sum_L2_square+=np.linalg.norm(grad_mu[:, :, 0], ord=2)**2
167         #print(sum_L2_square)
168         # A: energy derivative
169         A=(array_energy[istep]-array_energy[istep-1])/dtime
170         B= sum_L2_square
171         RES=A+B

```

```

172     print(RES)
173     return RES,A,B
174
175
176     # -----
177     def compute_residual_fd(dtime, array_energy,array_dfdc,array_c, kappa,istep):
178         sum_L2_square=0
179         grad_mu=np.zeros((3,1))
180         dfdc=array_dfdc[istep]
181         dfdc_1=array_dfdc[istep-1]
182         c=array_c[istep]
183         c_1=array_c[istep-1]
184         Nx=c.shape[0]
185         Ny=c.shape[1]
186         Nz=c.shape[2]
187         mu=np.zeros((Nx,Ny,Nz))
188         mu_p=np.zeros((Nx,Ny,Nz))
189         mu_m=np.zeros((Nx,Ny,Nz))
190         # compute \mu
191         for i in range(Nx):
192             for j in range(Ny):
193                 for l in range(Nz):
194                     mu_p[i,j,l]=dfdc[i,j,l]-kappa*laplacian_2D(c,i,j,l,dx_s,dy_s,dz_s)
195                     mu_m[i,j,l]=dfdc_1[i,j,l]-kappa*laplacian_2D(c_1,i,j,l,dx_s,dy_s,dz_s)
196         mu=(mu_p+ mu_m)/2
197         # compute Nabla(\mu) at each point grid and the associated L2 norm **2
198         for i in range(Nx):
199             for j in range(Ny):
200                 for l in range(Nz):
201                     grad_mu_x,grad_mu_y,grad_mu_z=gradient(mu,i,j,l,dx_s,dy_s,dz_s)
202                     grad_mu=np.array([grad_mu_x,grad_mu_y,grad_mu_z])
203                     sum_L2_square+=np.linalg.norm(grad_mu, ord=2)**2
204         # A: energy derivative
205         A=(array_energy[istep]-array_energy[istep-1])/dtime
206         B=sum_L2_square
207         RES=A+B
208
209     return RES, A, B

```

In [3]:

```

1  #-----
2  # compute residual at each time step (if istep >1), using Fourier space
3  def compute_residual_fft_test(array_energy,dtime, dfdck,ck,istep,k,k2, kappa,Nx,Ny,Nz):
4      Nx=c.shape[0]
5      Ny=c.shape[1]
6      Nz=c.shape[2]
7
8      # compute \mu
9      mu=dfdck+kappa*(k2)*ck
10     mu=np.real(np.fft.ifft(mu))
11
12     sum_L2_square=0
13     for i in range(Nx):
14         for j in range(Ny):
15             for l in range(Nz):
16                 grad_mu_x,grad_mu_y,grad_mu_z=gradient(mu,i,j,l,dx_s,dy_s,dz_s)
17                 grad_mu=np.array([grad_mu_x,grad_mu_y,grad_mu_z])
18                 sum_L2_square+=np.linalg.norm(grad_mu, ord=2)**2
19     # A: energy derivative
20     A=(array_energy[istep]-array_energy[istep-1])/dtime
21     B=sum_L2_square
22     RES=A+B
23     print(RES)
24     """
25     grad_mu=(1j*k[0])*mu +(1j*k[1])*mu+(1j*k[2])*mu
26     print( grad_mu.shape)
27     sum_L2_square=np.linalg.norm(grad_mu[:, :, 0], ord=2)**2
28     print(sum_L2_square)
29     # compute Nabla(\mu) at each point grid and the associated L2 norm **2
30     for i in range(Nx):
31         for j in range(Ny):
32             for l in range(Nz):
33                 grad_mu=np.array([ (1j*k[0,i,j,l])*mu[i,j,l] , (1j*k[1,i,j,l])*mu[i,j,l] , (1j*k[2,i,j,l])*mu[i,j,l])
34                 grad_mu=np.real(np.fft.ifft(grad_mu))
35                 sum_L2_square+=np.linalg.norm(grad_mu, ord=2)**2
36     #print(sum_L2_square)
37     # A: energy derivative
38     A=(array_energy[istep]-array_energy[istep-1])/dtime
39     B= sum_L2_square
40     RES=A+B
41     print(RES)
42     """

```


43	<code>return RES,A,B</code>
----	-----------------------------

In [4]:

1	<code>#RES, A, B=compute_residual_fft_test(array_energy,dtime, array_df_dc_k[nstep-1],ck,istep,k, k2,grad_coef,Nx,</code>
---	---

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In [5]:

```

1  # compute ND grad (Ny=Nz=1)
2  def gradient_ND(c,dx,dy,dz):
3      Nx=c.shape[0]
4      Ny=c.shape[1]
5      Nz=c.shape[2]
6      count_arr = np.bincount(np.array([Nx,Ny,Nz]))
7      dim=3-count_arr[1] # return problem dimension: 1D, 2D or 3D
8      grad_ND=np.zeros((3,Nx,Ny,Nz))
9      grad_ND_x=np.zeros((Nx,Ny,Nz))
10     grad_ND_y=np.zeros((Nx,Ny,Nz))
11     grad_ND_z=np.zeros((Nx,Ny,Nz))
12
13
14     if (dim==1):
15         for i in range(1,Nx):
16             grad_ND_x[i,0,0]=(c[i,0,0]-c[i-1,0,0])/(dx)
17
18     elif (dim==2): # for simplification, we suppose that Nz=1
19         for i in range(1,Nx):
20             for j in range(1,Ny):
21                 grad_ND_x[i,j,0]=(c[i,j,0]-c[i-1,j,0])/(dx)
22                 grad_ND_y[i,j,0]=(c[i,j,0]-c[i,j-1,0])/(dy)
23
24     else: # 3D
25         for i in range(1,Nx):
26             for j in range(1,Ny):
27                 for l in range(1,Nz):
28                     grad_ND_x[i,j,l]=(c[i,j,0]-c[i-1,j,0])/(dx)
29                     grad_ND_y[i,j,l]=(c[i,j,0]-c[i,j-1,0])/(dy)
30                     grad_ND_z[i,j,l]=(c[i,j,l]-c[i,j,l-1])/(dz)
31
32     grad_ND[0,:,:]=grad_ND_x
33     grad_ND[1,:,:]=grad_ND_y
34     grad_ND[2,:,:]=grad_ND_z
35
36     return grad_ND
37

```

In [6]:

```

1 @jit(nopython=True)
2 def prepar_fft(Nx,dx,Ny,dy,Nz,dz,opt):
3     """
4     Compute spatial frequency term and derivative
5     """
6     # variable initialisation
7     lin_x=np.zeros(Nx)
8     lin_y=np.zeros(Ny)
9     lin_z=np.zeros(Nz)
10
11     k=np.zeros((3,Nx,Ny,Nz))
12     k2=np.zeros((Nx,Ny,Nz))
13     k4=np.zeros((Nx,Ny,Nz))
14
15     """
16     # Method 1 to compute k (3D)
17     if (Nx % 2) == 1 : # = number odd if remainders is one
18         lin_x[:int((Nx-1)/2.0+1)]=np.arange(0, int((Nx-1)/2.0+1), 1)*2*np.pi/(Nx*dx)
19         lin_x[int((Nx-1)/2.0+1):]=np.arange(int(-(Nx+1)/2.0 +1), 0, 1)*2*np.pi/(Nx*dx)
20     if (Ny % 2) == 1 :
21         lin_y[:int((Ny-1)/2.0+1)]=np.arange(0, int((Ny-1)/2.0+1), 1)*2*np.pi/(Ny*dy)
22         lin_y[int((Ny-1)/2.0+1):]=np.arange(int(-(Ny+1)/2.0 +1), 0, 1)*2*np.pi/(Ny*dy)
23     if (Nz % 2) == 1 :
24         lin_z[:int((Nz-1)/2.0+1)]=np.arange(0, int((Nz-1)/2.0+1), 1)*2*np.pi/(Nz*dz)
25         lin_z[int((Nz-1)/2.0+1):]=np.arange(int(-(Nz+1)/2.0 +1), 0, 1)*2*np.pi/(Nz*dz)
26     if (Nx % 2) == 0 : # = number even if remainders is zero
27         lin_x[0:int(Nx/2.0)]=np.arange(0, int(Nx/2.0), 1)*2*np.pi/(Nx*dx)
28         lin_x[int(Nx/2.0 + 1):]=np.arange(int(-Nx/2.0 + 1), 0, 1)*2*np.pi/(Nx*dx)
29     if (Ny % 2) == 0 :
30         lin_y[0:int(Ny/2.0)]=np.arange(0, int(Ny/2.0), 1)*2*np.pi/(Ny*dy)
31         lin_y[int(Ny/2.0 + 1):]=np.arange(int(-Ny/2.0 + 1), 0, 1)*2*np.pi/(Ny*dy)
32     if (Nz % 2) == 0 :
33         lin_z[0:int(Nz/2.0)]=np.arange(0, int(Nz/2.0), 1)*2*np.pi/(Nz*dz)
34         lin_z[int(Nz/2.0 + 1):]=np.arange(int(-Nz/2.0 + 1), 0, 1)*2*np.pi/(Nz*dz)
35     # grid
36     for i in range(Nx):
37         for j in range(Ny):
38             for l in range(Nz):
39                 k[0,i,j,l]= lin_x[i]
40                 k[1,i,j,l]= lin_y[j]
41                 k[2,i,j,l]= lin_z[l]
42     """

```

```

43  # Method 2 to compute k
44  Lx=Nx*dx
45  x=np.linspace(-0.5*Lx+dx,0.5*Lx,Nx)
46
47  Ly=Ny*dy
48  y=np.linspace(-0.5*Ly+dy,0.5*Ly,Ny)
49
50  Lz=Nz*dz
51  z=np.linspace(-0.5*Lz+dz,0.5*Lz,Nz)
52
53
54  xx=2*np.pi/Lx*np.concatenate((np.arange(0,Nx/2+1), np.arange(-Nx/2+1,0)),axis=0)
55  yy=2*np.pi/Ly*np.concatenate((np.arange(0,Ny/2+1), np.arange(-Ny/2+1,0)),axis=0)
56  zz=2*np.pi/Lz*np.concatenate((np.arange(0,Nz/2+1), np.arange(-Nz/2+1,0)),axis=0)
57
58  for i in range(Nx):
59      for j in range(Ny):
60          for l in range(Nz):
61              k[0,i,j,l]= xx[i]
62              k[1,i,j,l]= yy[j]
63              k[2,i,j,l]= zz[l]
64
65  k2=k[0]**2+k[1]**2+k[2]**2
66
67  k4=k2**2
68
69  return k,k2,k4,x,y,z
70 def plot_micro(c,opt,ttime):
71     # 1D case
72     if (opt=='1D'):
73         plt.plot(c[:, :, 0])
74         plt.xlabel('x')
75         plt.ylabel('concentration')
76         plt.title('initial concentration')
77
78     else:
79         # 2D or 3D cases-----
80         import sys
81         grid = pv.UniformGrid()
82         grid.spacing=np.array([dx,dx,dx])*1E9
83         grid.dimensions = np.array([Nx,Ny,Nz])#+1
84         grid.point_arrays['c'] = np.transpose(np.resize(c,[Nx,Ny,Nz])).flatten()
85

```

```
86
87     # Set a custom position and size
88     sargs = dict(fmt="%.1f", color='black')
89
90     p = pv.Plotter()
91     pv.set_plot_theme("ParaView")
92     p.set_background("white")
93     p.add_mesh(grid, show_scalar_bar=False, label='title')
94     p.add_scalar_bar('Concentration', color='black', label_font_size=12, width=0.1, height=0.7, position
95     p.add_text('title')
96     p.show_bounds(all_edges=True, xlabel="x [nm]", ylabel="y [nm]", zlabel="z [nm]", color='black')
97     p.add_title('title')
98     p.camera_position = [1, 1, 1]
99     if (opt=='2D'):
100         if (ttime==0):
101             p.show(screenshot='Initial microstructure.png', cpos="xy") # cpos="xy" in case of 2D (Nz=1)
102         else:
103             p.show(screenshot='Microstructure at dimensionless time '+str('{0:.2f}'.format(ttime)) + '.
104     else:
105         if (ttime==0):
106             p.show(screenshot='Initial microstructure.png') # 3D plot
107             p.show(screenshot='Microstructure at dimensionless time '+str('{0:.2f}'.format(ttime)) + '.

```

In [7]:

```

1  # plot micro
2  def plot_micro(c,opt,ttime):
3      # 1D case
4      if (opt=='1D'):
5          plt.plot(c[:, :, 0])
6          plt.xlabel('x')
7          plt.ylabel('concentration')
8          plt.title('initial concentration')
9
10     else:
11         # 2D or 3D cases-----
12         import sys
13         grid = pv.UniformGrid()
14         grid.spacing=np.array([dx,dx,dx])*1E9
15         grid.dimensions = np.array([Nx,Ny,Nz])#+1
16         grid.point_arrays[r'c'] = np.transpose(np.resize(c,[Nx,Ny,Nz])).flatten()
17
18
19         # Set a custom position and size
20         sargs = dict(fmt="%.1f", color='black')
21
22         p = pv.Plotter()
23         pv.set_plot_theme("ParaView")
24         p.set_background("white")
25         p.add_mesh(grid,show_scalar_bar=False,label='title')
26         p.add_scalar_bar('Concentration', color='black',label_font_size=12, width=0.1, height=0.7, position='right')
27         if (ttime==0):
28             p.add_text('Initial Microstructure ',position='upper_edge',color='black',font='times',font_size=12)
29         else:
30             p.add_text('Microstructure at dimensionless time '+str('{0:.2f}'.format(ttime) ),color='black',font='times',font_size=12)
31
32         p.show_bounds(all_edges=True,font_size=24,bold=True, xlabel="X [nm]",ylabel="Y [nm]",zlabel="Z [nm]")
33         #p.add_title('Microstructure at dimensionless time '+str('{0:.2f}'.format(ttime) ))
34         p.camera_position = [1, 1, 1]
35         if (opt=='2D'):
36             if (ttime==0):
37                 p.show(screenshot='Initial microstructure.png',cpos="xy") # cpos="xy" in case of 2D (Nz=1)
38             else:
39                 p.show(screenshot='Microstructure at dimensionless time '+str('{0:.2f}'.format(ttime) )+ '.png')
40         elif (opt=='3D'):
41             if (ttime==0):
42                 p.show(screenshot='Initial microstructure.png') # cpos="xy" in case of 2D (Nz=1)

```

```

43         else:
44             p.show(screenshot='Microstructure at dimensionless time '+str('{0:.2f}'.format(ttime)) + '.|
45
46             p.close()

```

Input Data

In [8]:

```

1  Nx=64 ; Ny=64; Nz=64
2
3  # spacing
4  dx=10e-10 # [m]
5  dy=10e-10 # [m]
6  dz=10e-10 # [m]
7
8  # convert into adimensional grid
9  dx_s=dx/dx
10 dy_s=dy/dy
11 dz_s=dz/dz
12

```

Time stepping

Remind: This adaptive time-stepping method considers the time derivative of the total energy

$$\Delta t = \max(\Delta t_{min}, \frac{\Delta t_{max}}{\sqrt{1+\alpha|e'(t)|^2}})$$

In [9]:

```

1  array_alpha=[10]
2  array_dt_min= [0.1]
3  array_dt_max=[2]
4

```

```

In [10]: 1 c0=0.4 #initial microstructure
          2 mobility =1.0
          3 coefA = 1.0
          4 grad_coef=0.5
          5
          6
          7 # start simulation
          8 t_start = time.time()
          9
         10 #initialize microstructure
         11 c= micro_ch_pre(Nx,Ny,Nz,c0)
         12 c0_save=c
         13
         14 """
         15 #uncomment if necessary
         16 # For comparison purposes: load the same initial microstructure as Matlab
         17 # retrieving data from file.
         18 loaded_arr = np.loadtxt("3D_micro_init.txt")
         19
         20 # This loadedArr is a 2D array, therefore
         21 # we need to convert it to the original
         22 # array shape.reshaping to get original
         23 # matrice with original shape.
         24 load_original_arr = loaded_arr.reshape(
         25 loaded_arr.shape[0], loaded_arr.shape[1] // c.shape[2], c.shape[2])
         26
         27 c=load_original_arr
         28 """

```

```

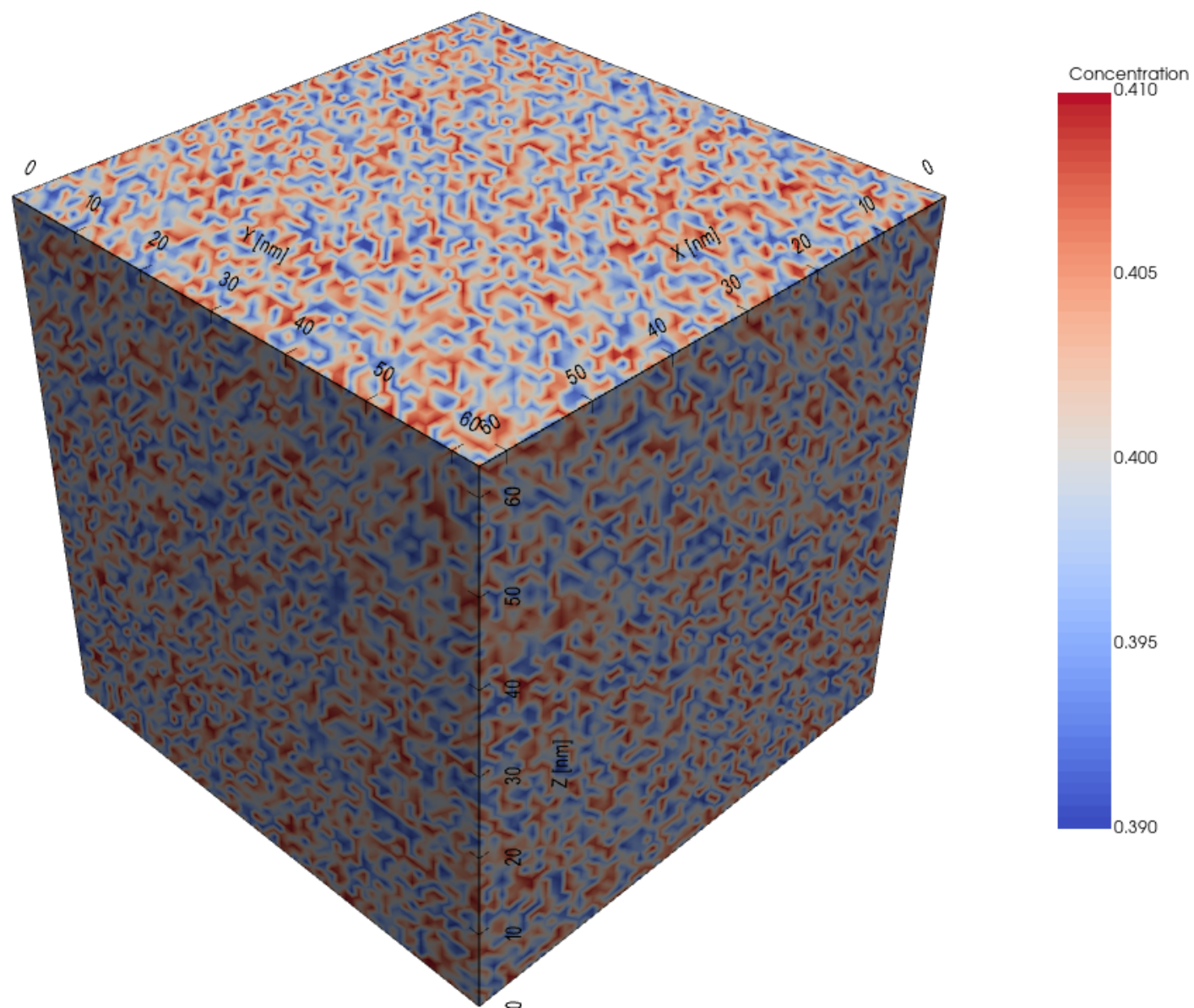
Out[10]: '\n#uncomment if necessary\n# For comparison purposes: load the same initial microstructure as Matlab\n# retrieving data from file.\nloaded_arr = np.loadtxt("3D_micro_init.txt")\n\n# This loadedArr is a 2D array, therefore\n# we need to convert it to the original\n# array shape.reshaping to get original\n# matrice with original shape.\nload_original_arr = loaded_arr.reshape(\nloaded_arr.shape[0], loaded_arr.shape[1] // c.shape[2], c.shape[2])\n\nc=load_original_arr\n'

```

plot initial microstructure


```
In [11]: 1 ttime=0  
2 plot_micro(c0_save,'3D',ttime)  
3
```

Initial Microstructure



Loop on time steps

```

In [12]: 1 # Loop to change time step (dtime) and compute associated energy dissipation for CH equation (at each time
2         for index in range(len(array_dt_max)): # to change alpha (or dtime_min or dtime_max) value in adaptive t
3             c=c0_save # start with the same microstructure
4             alpha=array_alpha[0]
5             dt_max=array_dt_max[0]
6             dt_min=array_dt_min[0]
7             # time step and constant values
8             ttime=0 # for each simulation
9             dtime=dt_min # dtime init
10            #-----
11            # time steps and print parameters
12            Nt=50 # trial
13            nstep= 1000 #int(round(Nt/dtime))
14            endstep=nstep # to change in loop ==> end simulation when some criteria are achieved (energy reaches a
15            nprint=100; # step to print
16            # set fourier coefficient
17            # compute the spatial frequency term from fft
18            k,k2,k4,x,y,z=prepar_fft(Nx,dx_s,Ny,dy_s,Nz,dz_s,opt="3d")
19            # initialize storage arrays
20
21            array_time=np.zeros(nstep)
22            array_dtime=np.zeros(nstep)
23
24            array_energy=np.zeros(nstep)
25            array_residus=np.zeros(nstep)
26            array_residus_fd=np.zeros(nstep)
27            array_energy_deriv=np.zeros(nstep) # default, computed in Fourier space
28            array_energy_deriv_fd=np.zeros(nstep) # "fd" computed by finite difference
29            array_energy_potentiel_grad=np.zeros(nstep)
30            array_energy_potentiel_grad_fd=np.zeros(nstep)
31            array_residual_deriv=np.zeros(nstep)
32
33
34            dfdc=np.zeros((Nx,Ny,Nz))
35            array_df_dc_k=np.zeros((nstep,Nx,Ny,Nz))
36            array_df_dc=np.zeros((nstep,Nx,Ny,Nz))
37            array_c_k=np.zeros((nstep,Nx,Ny,Nz))
38            array_c=np.zeros((nstep,Nx,Ny,Nz))
39
40            residual_deriv=1 # default value for the stop criteria
41            flag=0 # to stop simulation
42

```

```

43 t_start = time.time() # to compute CPU for each simulation
44 #-----
45 for istep in range(nstep):
46     #-----
47     ttime = ttime + dtime
48     array_dtime[istep]=dtime
49     array_time[istep]=ttime
50
51     # compute free energy
52     dfdc=free_energy(c)
53
54     dfdck=fft_(dfdc)
55     ck=fft_(c)
56
57     # Time integration
58     numer=dtime*mobility*k2*dfdck
59     denom = 1.0 + dtime*coefA*mobility*grad_coef*k4
60     ck =(ck-numer)/denom
61
62     c=np.real(iff_t_(ck))
63
64     # for small deviations
65     c[np.where(c >= 0.9999)]= 0.9999
66     c[np.where(c <= 0.00001)]=0.00001
67
68     energy=calculate_energy(Nx,Ny,Nz,c,grad_coef)
69     array_energy[istep]=energy
70     array_df_dc_k[istep]=dfdck
71     array_df_dc[istep]=dfdc
72     array_c_k[istep]=ck
73     array_c[istep]=c
74
75     if (math.fmod(istep,nprint)==0):
76         print()#plot_micro(c, '2D',ttime)
77
78     # adaptive time stepping
79     if (istep>0):
80         energy_derivative=(np.array(array_energy)[istep]-np.array(array_energy)[istep-1])/dtime
81
82         dtime=np.max([dt_min,dt_max/np.sqrt(1+alpha*(energy_derivative**2))])
83         #RE,energy_deriv,potentiel_grad=compute_residual_fft(array_energy,dtime, dfdck,ck,istep,k,k2, g
84         RE_fd,energy_deriv_fd,potentiel_grad_fd=compute_residual_fd(dtime, array_energy,array_df_dc,arr
85         #array_residus[istep]=RE

```

```

86     #array_energy_deriv[istep]=energy_deriv
87     #array_energy_potentiel_grad[istep]=potentiel_grad
88     array_energy_deriv_fd[istep]=energy_deriv_fd
89     array_energy_potentiel_grad_fd[istep]=potentiel_grad_fd
90     array_residus_fd[istep]=RE_fd
91
92     # criteria based on derivative of the energy residual
93     residual_deriv=(array_residus[istep]-array_residus[istep-1])/array_dtime[istep]
94     array_residual_deriv[istep]=residual_deriv
95     if (istep>1500): # (istep>200): by trial
96         residual_deriv_mean = np.mean([array_residual_deriv[j] for j in range(istep,istep-1
97         energy_deriv_mean = np.mean([array_energy_deriv[j] for j in range(istep,istep-100,-
98         #print(residual_deriv_mean)
99         if (np.abs(energy_deriv_mean)<0.01) :
100             print(istep,residual_deriv_mean)
101             flag=1
102             endstep=istep
103             break # break the for loop
104
105
106
107     #print('done istep: ', istep)
108     #-----
109     #-----
110     #----- Post processing -----
111     #-----
112     #-----
113     # plot energy evolution during spinodal decomposition
114     #-----
115
116     energ = np.array(array_energy)
117     rgb = np.random.rand(3,)
118     plt.plot(array_time[:endstep],energ[:endstep],label = r'$\alpha$='+str(alpha)+ ", "+ r'$\Delta T_{max}$'+s
119     plt.legend(loc='center left', bbox_to_anchor=(1, 0.5))
120     #plt.title('Energy evolution for dt='+str('{0:.2f}'.format(dtime)))
121     plt.xlabel('Dimensionless time')
122     plt.ylabel('Energy evolution')
123     #name='Energy evolution for different '+r'$\Delta T_{max}$' + ' values' + ".png"
124     #plt.savefig(name)
125     #plt.title('Energy evolution for different ' +r'$\Delta T_{max}$' + ' values')
126     plt.title('Energy evolution for optimum values of '+ r'$\Delta T_{max}$' + ', '+r'$\Delta T_{min}$' + ' and ' + r'$\
127     #plt.show()
128     #-----

```

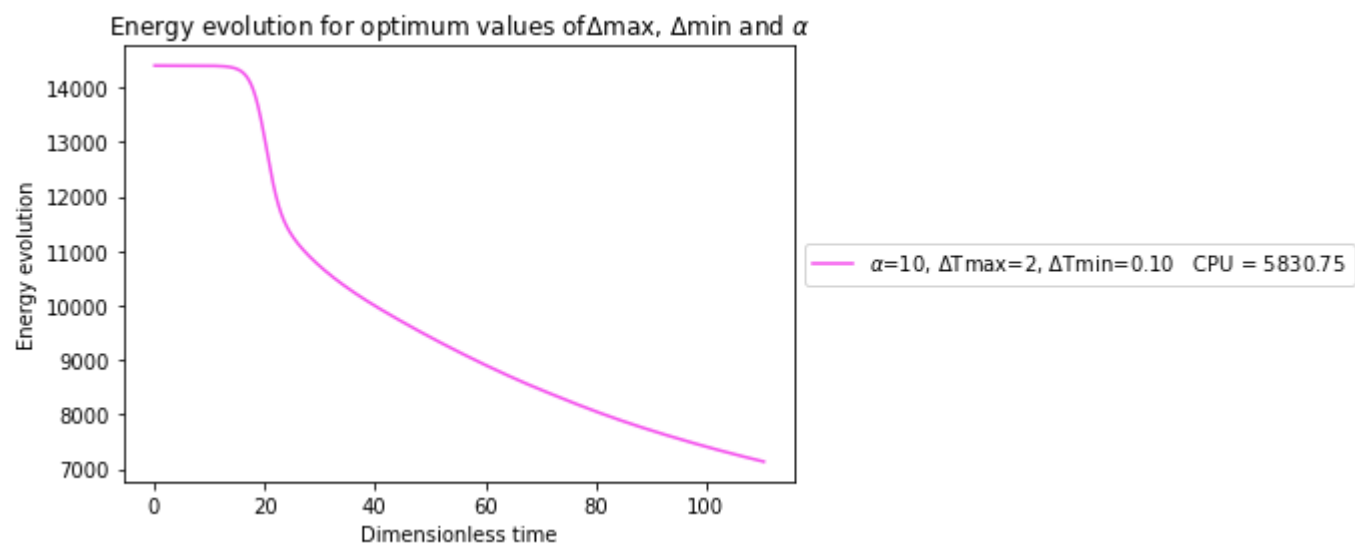
```

129 #plot solution evolution in space for (y=x) (2D)
130 #-----
131 """
132 #transform array_time (time steps sauvegarded) from list to array
133 array_time=np.array(array_time)
134
135 # return index of time of plot (here the end of simulation is chossen)
136 time_plot=int(np.array(np.where(abs(array_time - ttime)<0.001)))
137 print(time_plot)
138 #solution at the choosen instant
139 sol=array_c[time_plot]
140 # extract diagonal of the matrix "sol"
141 x_y_solution=np.diag(sol)
142 plt.xlabel('y=x')
143 plt.ylabel('Concentration at the end of simulation')
144 rgb = np.random.rand(3,)
145 plt.plot(x_y_solution,label = "dt="+str(dtime),color =rgb)
146 plt.legend(loc='center left', bbox_to_anchor=(1, 0.5))
147 plt.show()
148 """
149 #-----
150 # plot evolution of infinite norm versus time
151 #-----
152 """
153 infinite_norm=[]
154 array_c=np.array(array_c)
155 for i in range (1,nstep):
156     #compute infinite norm
157     c_n=array_c[i]
158     c_n__1=array_c[i-1]
159     infinite_norm.append(np.max(c_n-c_n__1))
160 infinite_norm=np.array( infinite_norm)
161 rgb = np.random.rand(3,)
162 plt.plot(array_time,infinite_norm,label = r'$\alpha$='+str(alpha)+ ", "+ r'$\Delta$Tmax='+str(dt_max)+
163 plt.xlabel('t$^{\{n\}}$')
164 plt.ylabel('||c$^{\{n\}}$ - c$^{\{n-1\}}$||$_{\{inf\}}$')
165 plt.legend(loc='center left', bbox_to_anchor=(1, 0.5))
166 plt.show()
167 """
168
169
170
171

```

```
172 print("simulation time : %s seconds ---" % (time.time() - t_start))
```

simulation time : 5830.80775809288 seconds ---

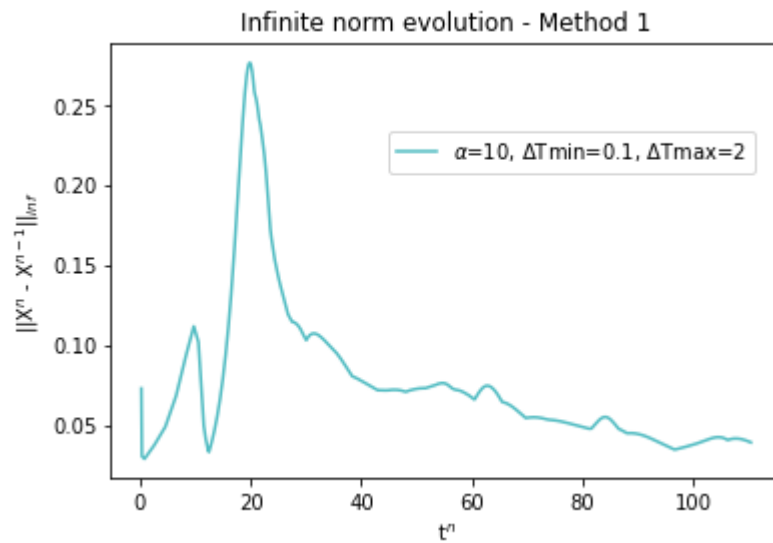


Postprocess

```

In [16]: 1 array_infinite_norm=[]
2 array_c=np.array(array_c)
3 for i in range (1,endstep):
4     #compute infinite norm
5     c_n=np.array(array_c[i])
6     c_n_1=np.array(array_c[i-1])
7     array_infinite_norm.append(infinite_norm(c_n-c_n_1))
8 array_infinite_norm=np.array(array_infinite_norm)
9 rgb = np.random.rand(3,)
10 plt.plot(array_time[1:endstep],array_infinite_norm,label=r'\alpha$=' +str(alpha)+ " , "+ r'\Delta$Tmin='+str(
11 plt.xlabel('t${n}$')
12 plt.ylabel('||X${n}$ - X${n-1}$||_{inf}$')
13 plt.title('Infinite norm evolution - Method 1')
14 plt.legend(loc='center left', bbox_to_anchor=(0.4, 0.75))
15 plt.show()

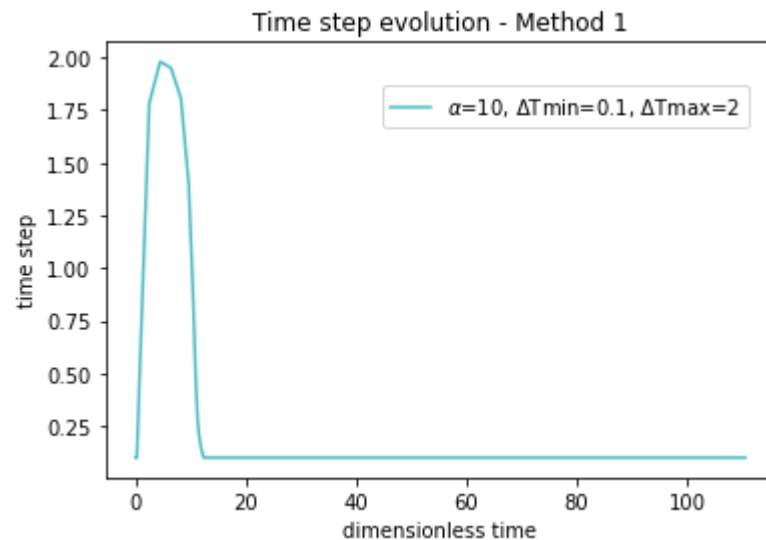
```




```
In [77]: 1 # save energy_vs_time in txt file : for comparative purposes
2 np.savetxt('energy_method_1.txt',energy[:endstep], fmt=' %.2f')
3 np.savetxt('time_method_1.txt',array_time[:endstep], fmt=' %.2f')
4
5 #energ_fd =np.loadtxt("energy_fd.txt", delimiter=" ", unpack=False)
```

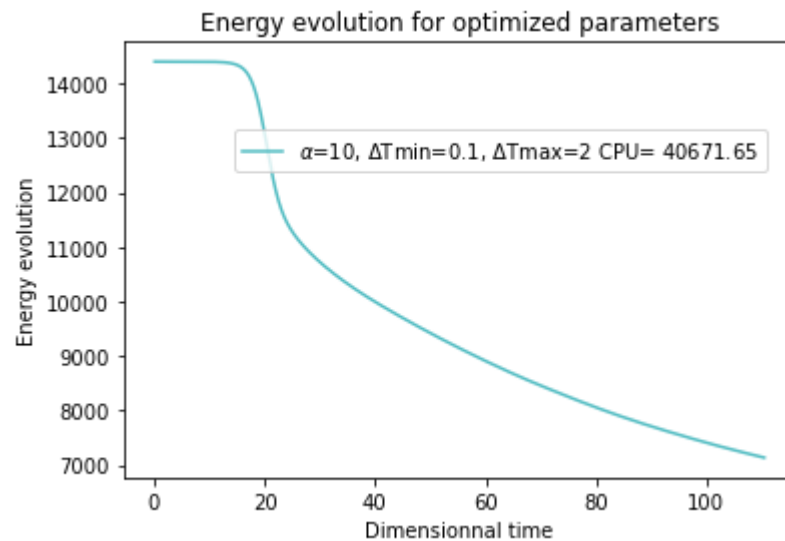
```
In [17]: 1 # plot dttime versus time evolution
2 plt.plot(array_time[:endstep],array_dtime[:endstep],label=r'$\alpha$=' +str(alpha)+ " , "+ r'$\Delta T_{min}$='
3 plt.legend(loc='center left', bbox_to_anchor=(0.4, 0.85))
4 plt.xlabel('dimensionless time')
5 plt.ylabel('time step')
6 plt.title('Time step evolution - Method 1')
```

Out[17]: Text(0.5, 1.0, 'Time step evolution - Method 1')



```
In [18]: 1 # Energy evolution
2 CPU=time.time() - t_start
3 energy = np.array(array_energy[:endstep])
4 plt.plot(array_time[:nstep],energy[:endstep],label=r'$\alpha$=' +str(alpha)+ " , "+ r'$\Delta$Tmin='+str(dt_
5 plt.legend(loc='center left', bbox_to_anchor=(0.15, 0.75))
6 plt.title('Energy evolution for optimized parameters')
7 plt.xlabel('Dimensionnal time')
8 plt.ylabel('Energy evolution')
9
```

Out[18]: Text(0, 0.5, 'Energy evolution')

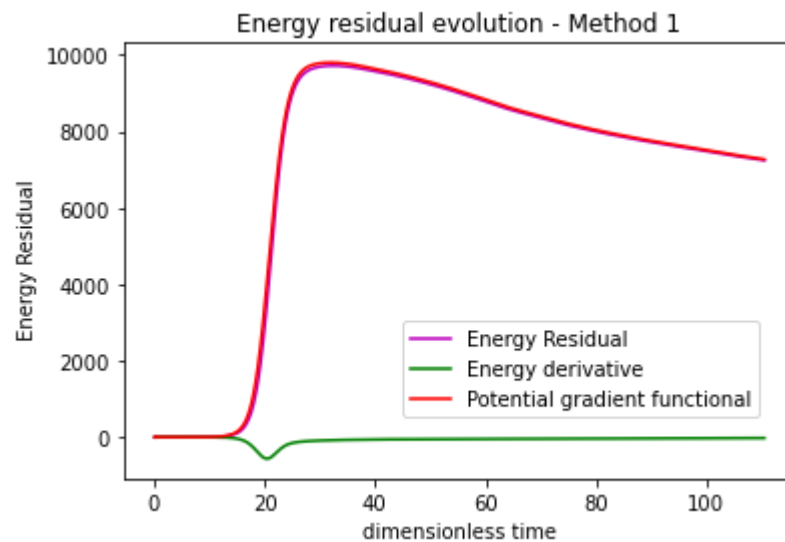


```

In [19]: 1 # plot dttime versus time evolution
          2 #plt.plot(array_time,array_residus, label='Residus (FFT)',c='b')
          3 plt.plot(array_time,array_residus_fd, label='Energy Residual',c='m')
          4 #plt.plot(array_time,array_energy_deriv, label='Energy derivative',c='g')
          5 plt.plot(array_time,array_energy_deriv_fd, label='Energy derivative',c='g')
          6 #plt.plot(array_time,array_energy_potentiel_grad, label='Potential gradient functional',c='r')
          7 plt.plot(array_time,array_energy_potentiel_grad_fd, label='Potential gradient functional',c='r')
          8 plt.legend(loc='center left', bbox_to_anchor=(0.4, 0.25))
          9 plt.xlabel('dimensionless time')
         10 plt.ylabel(' Energy Residual')
         11 plt.title('Energy residual evolution - Method 1')

```

Out[19]: Text(0.5, 1.0, 'Energy residual evolution - Method 1')



```
In [197]: 1 # save energy_vs_time in txt file : for comparative purposes
2 n_lines=energ.shape[0]
3 energy_time_adap=np.ones((n_lines,2))
4 energy_time_adap[:,0]=timesteps
5 energy_time_adap[:,1]=energ
6
7 np.savetxt('energy_versus_time_adaptive.txt', energy_time_adap, fmt='%.2f')
```

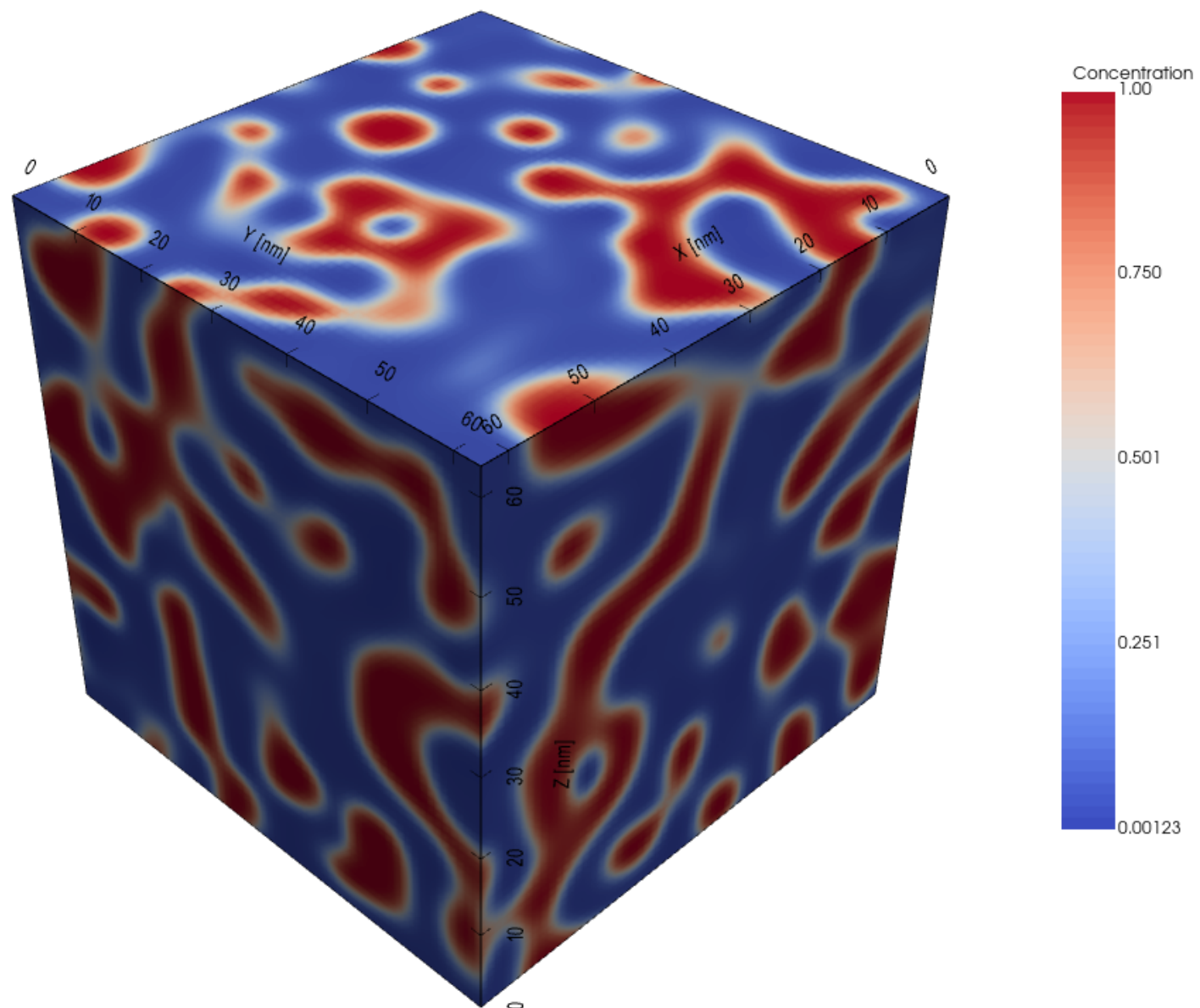
NameError Traceback (most recent call last)

```
<ipython-input-197-a356ed7c38e5> in <module>
      2 n_lines=energ.shape[0]
      3 energy_time_adap=np.ones((n_lines,2))
----> 4 energy_time_adap[:,0]=timesteps
      5 energy_time_adap[:,1]=energ
      6
```

NameError: name 'timesteps' is not defined

```
In [14]: 1 #plot actual microstructure  
2 plot_micro(c,'3D',ttime)
```

Microstructure at dimensionless time 110.43



```
In [15]: 1 #plot actual microstructure
          2 plot_mayavi(c,'3D',ttime)
          3 plot_m
```

```
-----
NameError                                Traceback (most recent call last)
<ipython-input-15-35b76a400f65> in <module>
      1 #plot actual microstructure
----> 2 plot_mayavi(c,'3D',ttime)

NameError: name 'plot_mayavi' is not defined
```

```
In [ ]: 1
```

