

Case Study # 5: Two-Species Diffusion-Diurnal Kinetics

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1 Problem Description

Chang et al. [1,2] have proposed approximate models to describe the chemical kinetics and transport phenomena associated with the dissociation of oxygen (O_2) into ozone (O_3) and monatomic oxygen (O) in the upper atmosphere. A one-dimensional version of such a model is considered here. The ambient oxygen concentration, c_3 , is constant, while the concentrations of the two minor species, O and O_3 , are $c_1(z, t)$ and $c_2(z, t)$, where z is the elevation above the earth's surface in km (here $30 \leq z \leq 50$) and t is time in seconds. Their transport is modeled using a reaction-diffusion equation,

$$\frac{\partial c_i}{\partial t} = \frac{\partial}{\partial z} \left[K(z) \frac{\partial c_i}{\partial z} \right] + R_i(\vec{c}, t) \quad i = 1, 2, 3. \quad (1)$$

The diffusive term is meant to represent the turbulent vertical transport with

$$K(z) = 10^{-8} \cdot \exp(z/5) \quad [\text{km}^2/\text{s}], \quad (2)$$

and the chemistry is described using the Chapman mechanism [2]. The reaction rates, $R_i(c, t)$, are given by

$$\begin{aligned} R_1(c_1, c_2, t) &= -k_1 c_1 c_3 - k_2 c_1 c_2 + 2k_3(t) c_3 + k_4(t) c_2 \\ R_2(c_1, c_2, t) &= k_1 c_1 c_3 - k_2 c_1 c_2 - k_4(t) c_2 \end{aligned} \quad (3)$$

with,

$$\begin{aligned} k_1 &= 1.63 \times 10^{16} \\ k_2 &= 4.66 \times 10^{16} \\ k_l &= \exp[a_l / \sin \omega t] \text{ if } \sin \omega t > 0, \text{ else } 0 \quad (l = 3, 4) \end{aligned}$$

and with $a_3 = 22.62$, $a_4 = 7.601$, and $\omega = \pi/43200$. This system is subject to the initial conditions,

$$\begin{aligned} c_1(z, 0) &= 10^6 \cdot \gamma(z) \\ c_2(z, 0) &= 10^{12} \cdot \gamma(z), \end{aligned} \quad (4)$$

where

$$\gamma(z) = 1 - \left(\frac{z-40}{10} \right)^2 + \frac{1}{2} \left(\frac{z-40}{10} \right)^4, \quad (5)$$

and a boundary condition of no flux at the top and bottom of the vertical atmospheric layer considered.

2 Numerical Solution Approach

To generate a system of ordinary differential equations, all the spatial derivatives in Equation 1 are replaced with centered finite differences. For the base case considered, the domain is discretized into $M = 50$ partitions, $\Delta z = 20/M$, and $z_j = 30 + j(\Delta z)$ for $0 \leq j \leq M$.

The function $c^i(z_j, t)$ can then be approximated as

$$\begin{aligned} \dot{c}_j^i &= (\Delta z)^{-2} [K_{j+1/2} c_{j+1}^i - (K_{j+1/2} + K_{j-1/2}) c_j^i + \\ &\quad K_{j-1/2} c_{j-1}^i] + R^i(\mathbf{c}, t), \end{aligned} \quad (6)$$

where $K_{j \pm 1/2} = K(30 + [j \pm 1/2] \Delta z)$. A system of $2M$ ODEs is then specified by setting $\mathbf{y}(t) = [c_1^1(t), c_1^2(t), c_2^1(t), c_2^2(t), \dots, c_M^1(t), c_M^2(t)]^T$, with boundary conditions $c_0^i = c_2^i$ and $c_{M-1}^i = c_{M+1}^i$. The two parabolic PDEs are then reduced to a system of $2M$ ODEs of the form $\dot{\mathbf{y}} = \mathbf{f}(t, \mathbf{y})$.

Two solvers are called from scipy version 0.11.0 to solve this system of ODEs. A stiff solver, "vode", is an implicit method based on the backward differentiation formulas, and a non-stiff solver, "dopri5", is an explicit Runge-Kutta method of order (4,5) are used. Both methods required

an absolute tolerance of $1\text{E-}1$ and a relative tolerance $1\text{E-}3$ for convergence.

3 Results Discussion

3.1 Comparison to Published Results

3.2 Sensitivity to Mesh Density

Solver	24 hours	10 days
dopri5	1025	10069
bdf	1318	13237

Table 1: Wall clock time, in seconds, to solve to $t = 24$ hours and $t = 10$ days

M	c_1	c_2
5	1.628E-2	1.648E-2
10	9.766E-3	9.829E-3
25	4.178E-3	4.110E-3
50	1.879E-3	1.811E-3
75	1.031E-3	9.806E-4
100	6.923E-4	6.098E-4

Table 2: NRMS of results at $t = 4$ hours compared to results at $M = 200$ (dopri5 solver)

M	c_1	c_2
5	1.654E-2	1.648E-2
10	9.796E-3	9.829E-3
25	4.091E-3	4.110E-3
50	1.802E-3	1.811E-3
75	9.436E-4	9.806E-4
100	5.982E-4	6.098E-4

Table 3: NRMS of results at $t = 4$ hours compared to results at $M = 200$ (bdf solver)

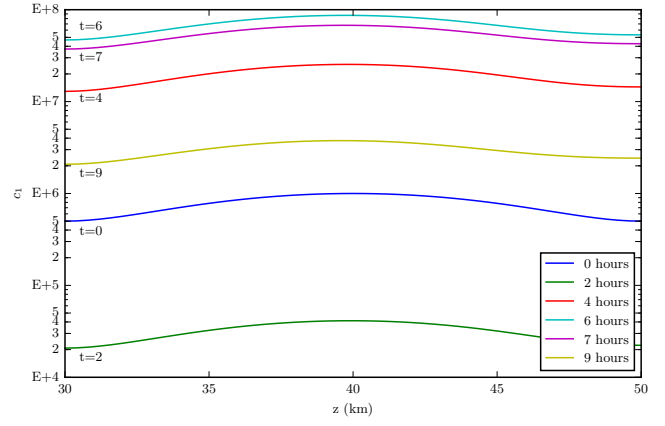


Fig. 1: c_1 vs. z at $t = 0, 2, 4, 6, 7$, and 9 hours

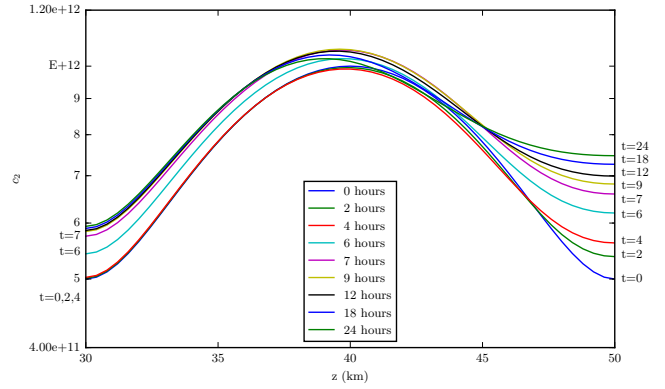


Fig. 2: c_2 vs. z at $t = 0, 2, 4, 6, 7, 9, 12, 18$, and 24 hours

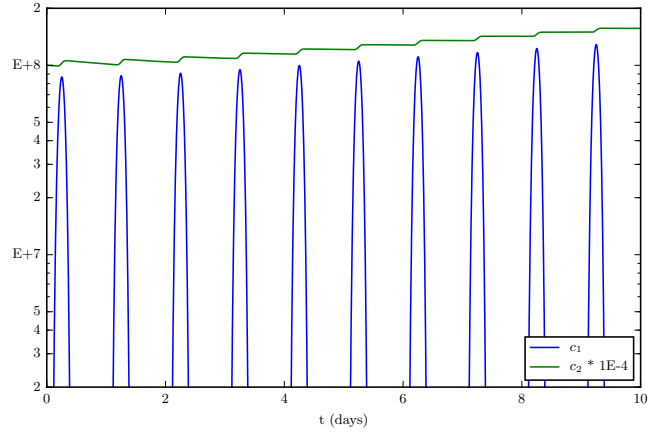


Fig. 3: c_1 and c_2 vs. time (from 0 to 10 days) at $z = 40$ km

4 Conclusion

References

- [1] Chang, J., Hindmarsh, A., and Madsen, N., 1974. "Simulation of chemical kinetics transport in the stratosphere". In *Stiff differential systems*. Springer, pp. 51–65.
- [2] Byrne, G. D., and Hindmarsh, A. C., 1987. "Stiff ode solvers: A review of current and coming attractions".

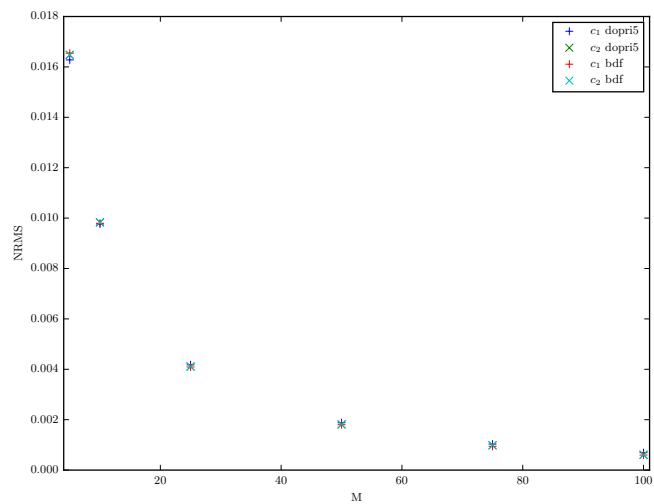


Fig. 4: NRMS for $M = 5, 10, 25, 50, 75$, and 100 compared against $M = 200$ for both solvers and c_1 and c_2

Appendix A: Python Code

```
1 import numpy as np
2 from scipy.integrate import ode
3 from time import clock
4 from PrettyPlots import *
5
6
7 def K(z):
8     zz = 30. + z * dz
9     return 1E-8 * np.exp(zz / 5.)
10
11
12 def gamma(z):
13     return 1. - ((z - 40.) / 10.) ** 2 + (1. / 2.) * ((z - 40.) / 10.) ** 4
14
15
16 def R(y_1, y_2, t):
17     '''
18     Find the reaction rates, R_1 and R_2, of the system at state c and time t.
19     '''
20
21     if np.sin(w * t) > 0.:
22         k_3 = np.exp(-a_3 / np.sin(w * t))
23         k_4 = np.exp(-a_4 / np.sin(w * t))
24     else:
25         k_3 = 0.
26         k_4 = 0.
27
28     R_1 = -k_1 * y_1 * y_3 - k_2 * y_1 * y_2 + 2. * k_3 * y_3 + k_4 * y_2
29     R_2 = +k_1 * y_1 * y_3 - k_2 * y_1 * y_2 - k_4 * y_2
30     return R_1, R_2
31
32
33 def system(t, y):
34     f = np.zeros(len(y))
35
36     R1, R2 = R(y[0], y[1], t)
37     l_p, l_m = 3. / 2., 1. / 2.
38
39     f[0] = (dz ** -2 * (K(l_p) * y[2] - (K(l_p) + K(l_m)) * y[0] + K(l_m) * y[2])) + R1)
40     f[1] = (dz ** -2 * (K(l_p) * y[3] - (K(l_p) + K(l_m)) * y[1] + K(l_m) * y[3])) + R2)
41
42     for i in range(1, M):
43         R1, R2 = R(y[2 * i], y[2 * i + 1], t)
44         l_p, l_m = 1. * i + 3. / 2., 1. * i + 1. / 2.
45
46         f[2 * i] = (dz ** -2 * (K(l_p) * y[2 * i + 2] -
47                                (K(l_p) + K(l_m)) * y[2 * i] + K(l_m) * y[2 * i - 2])) + R1)
48         f[2 * i + 1] = (dz ** -2 * (K(l_p) * y[2 * i + 3] -
49                                (K(l_p) + K(l_m)) * y[2 * i + 1] + K(l_m) * y[2 * i - 1])) + R2)
50
51     R1, R2 = R(y[2 * M], y[2 * M + 1], t)
52     l_p, l_m = 1. * M + 1. / 2., 1. * M - 1. / 2.
53
54     f[2 * M] = (dz ** -2 * (K(l_p) * y[2 * M - 2] -
55                            (K(l_p) + K(l_m)) * y[2 * M] + K(l_m) * y[2 * M - 2])) + R1)
56     f[2 * M + 1] = (dz ** -2 * (K(l_p) * y[2 * M - 1] -
57                            (K(l_p) + K(l_m)) * y[2 * M + 1] + K(l_m) * y[2 * M - 1])) + R2)
58
59     return f
60
61
62 def solve(solver, c, time, integrator):
63     # Create result arrays
64     c1, c2, c1_40km, c2_40km, t = [], [], [], [], []
65
66     start_time = clock()
67     for i in range(0, len(time) - 1):
```

```

68     # Initial and final time
69     t_0 = time[i]
70     t_f = time[i + 1]
71
72     # Solver setup
73     sol = []
74     solver.set_initial_value(c, t_0)
75     while solver.successful() and solver.t < t_f:
76         solver.integrate(solver.t + dt)
77         sol.append(solver.y)
78
79     # keep time history for 40km point
80     one, two = sol[-1][0::2], sol[-1][1::2]
81     mid_one, mid_two = one[M / 2], two[M / 2]
82     c1_40km.append(mid_one), c2_40km.append(mid_two)
83     t.append(solver.t)
84
85     print "{0:03.2f}%".format(100. * solver.t / time[-1])
86
87     # Save c1, c2 solutions
88     c1.append(one), c2.append(two)
89
90     #Update initial conditions for next iteration
91     c = sol[-1]
92
93     elapsed_time = clock() - start_time
94     print(elapsed_time, "seconds process time")
95
96     output = [c1, c2, c1_40km, c2_40km, t]
97     return output
98
99
100 def save_variables(name, z, c1, c2, t, c1_40km, c2_40km):
101     try:
102         os.mkdir('data')
103     except Exception:
104         pass
105
106     try:
107         os.mkdir('data/' + name)
108     except Exception:
109         pass
110
111     np.savetxt('data/' + name + '/z.csv', z)
112     np.savetxt('data/' + name + '/c1.csv', c1)
113     np.savetxt('data/' + name + '/c2.csv', c2)
114     np.savetxt('data/' + name + '/t.csv', t)
115     np.savetxt('data/' + name + '/c1_40km.csv', c1_40km)
116     np.savetxt('data/' + name + '/c2_40km.csv', c2_40km)
117
118
119 def load_variables(name):
120     z = np.loadtxt('data/' + name + '/z.csv')
121     c1 = np.loadtxt('data/' + name + '/c1.csv')
122     c2 = np.loadtxt('data/' + name + '/c2.csv')
123     t = np.loadtxt('data/' + name + '/t.csv')
124     c1_40km = np.loadtxt('data/' + name + '/c1_40km.csv')
125     c2_40km = np.loadtxt('data/' + name + '/c2_40km.csv')
126
127     return z, c1, c2, t, c1_40km, c2_40km
128
129
130 def run_trials(z, integrators, times, M):
131     # Set up ODE solver
132     for integrator in integrators:
133         if integrator == 'dop853' or integrator == 'dopri5':
134             solver = ode(system)
135             solver.set_integrator(integrator, atol=1E-1, rtol=1E-3)
136             elif integrator == 'bdf':

```

```

137     solver = ode(system)
138     solver.set_integrator('vode', method=integrator, atol=1E-1, rtol=1E-3, nsteps=2000)
139
140     name = integrator + ' ' + str(times[-1]) + ' ' + str(M)
141     try:
142         z, c1, c2, t, c1_40km, c2_40km = load_variables(name)
143         print "Loaded data for: " + name
144     except:
145         print "Starting solver: ", integrator, "with times", times
146         c1, c2, c1_40km, c2_40km, t = solve(solver, c, times, integrator)
147         save_variables(name, z, c1, c2, t, c1_40km, c2_40km)
148
149     # And plot some things
150     if times[-1] == 86400.0:
151         labels = [str(int(time / 3600.)) + " hours" for time in times[1:]]
152         plot_c1(z, c, c1, labels, name)
153         plot_c2(z, c, c2, labels, name)
154     elif times[-1] == 864000.0:
155         plot_40km(t, c1_40km, c2_40km, name)
156
157
158 def sensitivity_analysis(integrators, times, meshes):
159     plt.figure()
160     for integrator in integrators:
161         z_M, c1_M, c2_M = [], [], []
162
163         for M in meshes:
164             name = integrator + ' ' + str(times[-1]) + ' ' + str(M)
165             try:
166                 z, c1, c2, _, _, _ = load_variables(name)
167             except Exception:
168                 print Exception
169             z_M.append(list(z))
170             c1_M.append(list(c1[-1]))
171             c2_M.append(list(c2[-1]))
172
173         best_z = z_M[-1]
174         best_c1, best_c2 = c1_M[-1], c2_M[-1]
175         NRMS1, NRMS2 = [], []
176         for j, mesh in enumerate(z_M):
177             if j + 1 == len(z_M): break # RMS with yourself is silly
178             best1, best2, curr1, curr2 = [], [], [], []
179             for i, element in enumerate(best_z):
180                 if element in mesh:
181                     best1.append(best_c1[i])
182                     best2.append(best_c2[i])
183                     curr1.append(c1_M[j][mesh.index(element)])
184                     curr2.append(c2_M[j][mesh.index(element)])
185
186             best1, best2 = np.array(best1), np.array(best2)
187             curr1, curr2 = np.array(curr1), np.array(curr2)
188
189             err1, err2 = curr1 - best1, curr2 - best2
190             NRMS1.append(np.sqrt(np.mean(np.square(err1)))/(max(best1) - min(best1)))
191             NRMS2.append(np.sqrt(np.mean(np.square(err2)))/(max(best2) - min(best2)))
192             # print meshes[j], NRMS1, NRMS2
193
194         x = [mesh for mesh in meshes][0:-1]
195         plt.plot(x, NRMS1, '+', label='$c_1$ ' + integrator)
196         plt.plot(x, NRMS2, 'x', label='$c_2$ ' + integrator)
197
198     plt.ylabel('NRMS')
199     plt.xlabel('M')
200     plt.xlim([meshes[0] - 1, meshes[-2] + 1])
201     plt.legend()
202     save_name = str(meshes) + '.pdf'
203     save_plot(save_name)
204
205

```

```

206 # Basic problem parameters
207 y_3 = 3.7E16          # Concentration of O_2 (constant)
208 k_1 = 1.63E-16        # Reaction rate [O + O_2 -> O_3]
209 k_2 = 4.66E-16        # Reaction rate [O + O_3 -> 2 * O_2]
210 a_3 = 22.62           # Constant used in calculation of k_3
211 a_4 = 7.601           # Constant used in calculation of k_4
212 w = np.pi / 43200.   # Cycle (half a day) [1/sec]
213 dt = 60.
214
215 # Base Case
216 M = 50                # Number of sections
217 dz = 20. / M          # 20km divided by M subsections
218
219 # This generates the initial conditions
220 c = np.zeros(2 * (M + 1))
221 z = np.zeros(M + 1)
222 for i in range(0, M + 1):
223     z[i] = 30. + i * dz
224     c[2 * i] = 1E6 * gamma(z[i])
225     c[2 * i + 1] = 1E12 * gamma(z[i])
226
227 # Run the trials
228 integrators = ['dopri5', 'bdf']
229 times = 3600. * np.array([0., 2., 4., 6., 7., 9., 12., 18., 24.])
230 run_trials(z, integrators, times, M)
231
232 integrators = ['dopri5', 'bdf']
233 times = 3600. * np.array([0., 2., 4., 6., 7., 9., 12., 18., 240.])
234 run_trials(z, integrators, times, M)
235
236 # Mesh Analysis
237 meshes = [5, 10, 25, 50, 75, 100, 200]
238 for M in meshes:
239     dz = 20. / M          # 20km divided by M subsections
240
241     # This generates the initial conditions
242     c = np.zeros(2 * (M + 1))
243     z = np.zeros(M + 1)
244     for i in range(0, M + 1):
245         z[i] = 30. + i * dz
246         c[2 * i] = 1E6 * gamma(z[i])
247         c[2 * i + 1] = 1E12 * gamma(z[i])
248
249     # Time array
250     dt = 60.
251
252     integrators = ['dopri5', 'bdf']
253     times = 3600. * np.array([0., 2., 4.])
254     run_trials(z, integrators, times, M)
255
256 sensitivity_analysis(integrators, times, meshes)

```

Listing 1: Code to create solutions

```

1 import numpy as np
2 import matplotlib
3 matplotlib.use('TkAgg')
4 import matplotlib.pyplot as plt
5 import os
6
7 # Configure figures for production
8 WIDTH = 495.0 # the number latex spits out
9 FACTOR = 1.0 # the fraction of the width the figure should occupy
10 fig_width_pt = WIDTH * FACTOR
11
12 inches_per_pt = 1.0 / 72.27
13 golden_ratio = (np.sqrt(5) - 1.0) / 2.0 # because it looks good
14 fig_width_in = fig_width_pt * inches_per_pt # figure width in inches
15 fig_height_in = fig_width_in * golden_ratio # figure height in inches

```

```

16 fig_dims = [fig_width_in, fig_height_in] # fig dims as a list
17
18
19 def save_plot(save_name):
20     # Save plots
21     try:
22         os.mkdir('figures')
23     except Exception:
24         pass
25
26     plt.savefig('figures/' + save_name, bbox_inches='tight')
27     plt.close()
28
29
30 def plot_c1(z, initial, c1, labels, integrator):
31     plt.figure(figsize=fig_dims)
32     plt.plot(z, initial[0::2], label='0 hours')
33     for solution, label in zip(c1, labels):
34         if "12" in label:
35             break
36     plt.plot(z, solution, label=label)
37     plt.ylabel('$c_1$')
38     plt.xlabel('z (km)')
39     plt.yscale('log')
40     plt.ylim([1E4, 1E8])
41     plt.yticks([1E4, 2E4, 3E4, 4E4, 5E4,
42                1E5, 2E5, 3E5, 4E5, 5E5,
43                1E6, 2E6, 3E6, 4E6, 5E6,
44                1E7, 2E7, 3E7, 4E7, 5E7, 1E8],
45               ['E+4', '2', '3', '4', '5',
46                'E+5', '2', '3', '4', '5',
47                'E+6', '2', '3', '4', '5',
48                'E+7', '2', '3', '4', '5', 'E+8'])
49     plt.legend(loc='lower right')
50
51     plt.text(30.5, 1.5e+4, 't=2', fontsize=9, family='serif')
52     plt.text(30.5, 3.5e+5, 't=0', fontsize=9, family='serif')
53     plt.text(30.5, 1.5e+6, 't=9', fontsize=9, family='serif')
54     plt.text(30.5, 1.e+7, 't=4', fontsize=9, family='serif')
55     plt.text(30.5, 2.8e+7, 't=7', fontsize=9, family='serif')
56     plt.text(30.5, 6.e+7, 't=6', fontsize=9, family='serif')
57
58     save_name = integrator + ' c1.pdf'
59     save_plot(save_name)
60
61
62 def plot_c2(z, initial, c2, labels, integrator):
63     plt.figure(figsize=fig_dims)
64     plt.plot(z, initial[1::2], label='0 hours')
65     for solution, label in zip(c2, labels):
66         if "240" in label:
67             break
68     plt.plot(z, solution, label=label)
69     plt.ylabel('$c_2$')
70     plt.xlabel('z (km)')
71     plt.yscale('log')
72     plt.ylim([4.E11, 1.2E12])
73     plt.yticks([4E11, 5E11, 6E11, 7E11, 8E11, 9E11, 1E12, 1.2E12],
74               ['4.00e+11', '5', '6', '7', '8', '9', 'E+12', '1.20e+12'])
75     plt.legend(loc='best')
76
77     # Left side text
78     plt.text(28.25, 4.65e+11, 't=0,2,4', fontsize=9, family='serif')
79     plt.text(29, 5.4e+11, 't=6', fontsize=9, family='serif')
80     plt.text(29, 5.7e+11, 't=7', fontsize=9, family='serif')
81
82     # Right side text
83     plt.text(50.25, 4.95e+11, 't=0', fontsize=9, family='serif')
84     plt.text(50.25, 5.35e+11, 't=2', fontsize=9, family='serif')

```



```

85 plt.text(50.25, 5.6e+11, 't=4', fontsize=9, family='serif')
86 plt.text(50.25, 6.1e+11, 't=6', fontsize=9, family='serif')
87 plt.text(50.25, 6.4e+11, 't=7', fontsize=9, family='serif')
88 plt.text(50.25, 6.7e+11, 't=9', fontsize=9, family='serif')
89 plt.text(50.25, 7.0e+11, 't=12', fontsize=9, family='serif')
90 plt.text(50.25, 7.3e+11, 't=18', fontsize=9, family='serif')
91 plt.text(50.25, 7.6e+11, 't=24', fontsize=9, family='serif')
92
93 save_name = integrator + ' c2.pdf'
94 save_plot(save_name)
95
96
97 def plot_40km(t, c1_40km, c2_40km, integrator):
98     c2_40km_scaled = [1E-4 * val for val in c2_40km]
99     days = [val / 86400. for val in t]
100
101     plt.figure(figsize=fig_dims)
102     plt.plot(days, c1_40km, label='$c_1$')
103     plt.plot(days, c2_40km_scaled, label='$c_2$ * 1E-4')
104     plt.xlabel('t (days)')
105     plt.yscale('log')
106     plt.ylim([2.E6, 2E8])
107     plt.yticks([2E6, 3E6, 4E6, 5E6, 1E7, 2E7, 3E7, 4E7, 5E7, 1E8, 2E8],
108               ['2', '3', '4', '5', 'E+7', '2', '3', '4', '5', 'E+8', '2'])
109     plt.xlim([0, days[-1]])
110     plt.legend(loc='lower right')
111     save_name = integrator + ' time.pdf'
112     save_plot(save_name)

```

Listing 2: Code to generate pretty plots