

# Exercise 3, TFY4235 Computational physics

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## Introduction

This is an implementation of [1].

## Theory and implementation

The diffusion equation, can be written as

$$\Delta t \frac{\partial}{\partial t} C(z, t) = \Delta t \left( K(z) \frac{\partial^2}{\partial z^2} + \frac{dK(z)}{dz} \frac{\partial}{\partial z} \right) C(z, t) = \mathcal{D}C(z, t).$$

Discretizing the spatial part, and applying boundary conditions, gives

$$\Delta t \frac{\partial}{\partial t} C_n(t) = \mathcal{D}_{nm} C_n(t) + S_n(t),$$

where

$$\mathcal{D} = \begin{pmatrix} -4\alpha K_0 - 2\Gamma & 4\alpha K_0 & 0 & \dots & 0 \\ -\frac{\alpha}{2} K'_1 + 2\alpha K_1 & -4\alpha K_1 & \frac{\alpha}{2} K'_1 + 2\alpha K_1 & \dots & 0 \\ 0 & \ddots & \ddots & \ddots & 0 \\ 0 & \dots & -\frac{\alpha}{2} K'_{N-1} + 2\alpha K_{N-1} & -4\alpha K_{N-1} & \frac{\alpha}{2} K'_{N-1} + 2\alpha K_{N-1} \\ 0 & \dots & 0 & 4\alpha K_N & -4\alpha K_N \end{pmatrix},$$
$$S(t) = (2\Gamma C_{eq}(t) \quad 0 \quad \dots \quad 0)^T \quad \Gamma = 2 \frac{\alpha k_w \Delta z}{K_0} \left( K_0 - \frac{1}{2} \left( -\frac{3}{2} K_0 + 2K_1 - \frac{1}{2} K_2 \right) \right), \quad \alpha = \frac{\Delta t}{2\Delta z^2},$$
$$K'_n = K_{n+1} - K_{n-1}$$

The Cranck-Nichelson scheme then yields

$$C_n^{i+1} = C_n^i + \frac{1}{2} (\mathcal{D}_{nm} C_m^i + S_n^i) + \frac{1}{2} (\mathcal{D}_{nm} C_m^{i+1} + S_n^{i+1}),$$

so the equation to be solved to get the next timestep is

$$A_{nm} C_m^{i+1} = V_n^i,$$
$$V_n^i = \left( \delta_{nm} + \frac{1}{2} \mathcal{D}_{nm} \right) C_m^i + \frac{1}{2} (S_n^i + S_n^{i+1}), \quad A_{mn} = \left( \delta_{nm} - \frac{1}{2} \mathcal{D}_{nm} \right)$$

The implementation of this system of equation uses SciPy's sparse matrix library. After creating sparse realizations of  $A$ , SciPy's `splu` is used to generate the LU decomposition LU of  $A$ . This is an object with methods such as `.solve()`, which utilizes the LU decomposition. The wrapper `simulate` then loops over  $N_t - 1$  steps, using `solve(V) = lambda V: LU.solve(V)`, where  $V$  is as given above.

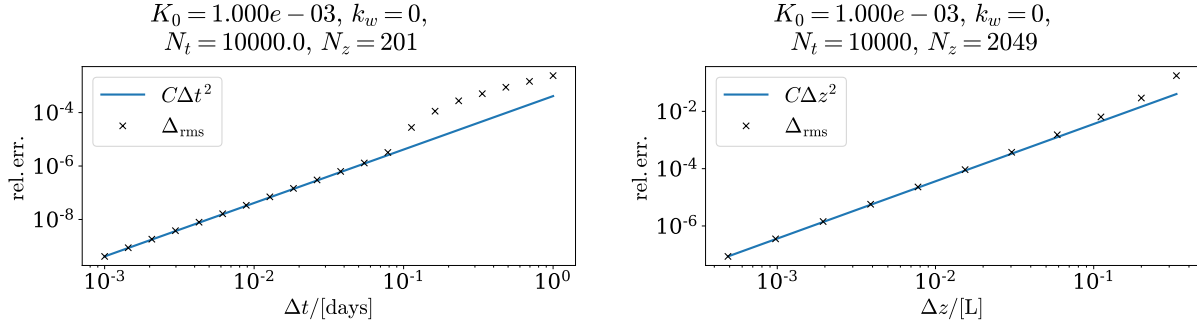


Figure 1: Error, measured as the root mean square deviation from a reference value, after 1 day simulation of an Gaussian initial concentration.

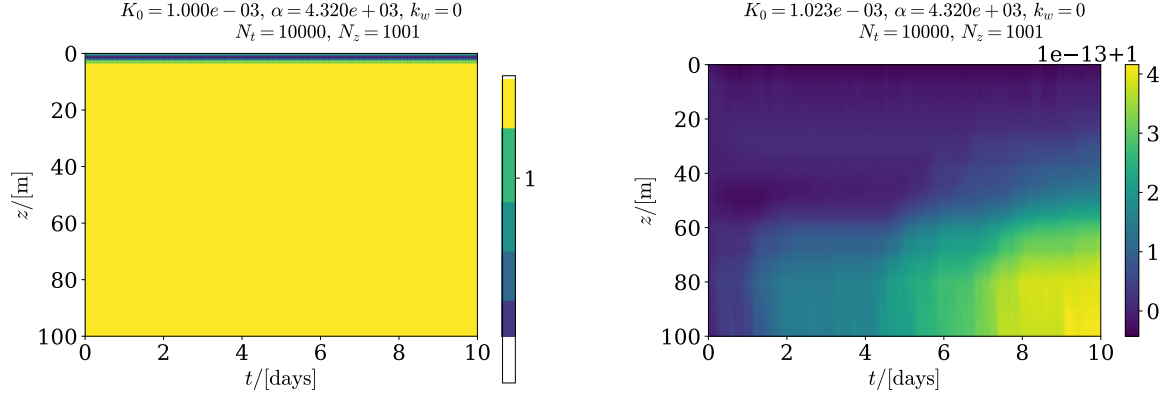


Figure 2: The time evolution of an constant concentration. The system on the left has a constant  $K(z)$ , while system of the right has an oscillatory  $K$ . The largest deviation of the system is of order  $10^{-15}$  and  $10^{-13}$ , respectively.

## Tests

Make sure the implementation gives good answers, it is compared to known solutions. The method used in this implementation has quadratic convergence, both in time and space. A convergence test was implemented for a simple test case, to check this. Figure 1 shows the result of this.

A constant concentration of  $\text{CO}_2$  should remain constant, regardless of  $K(z)$ , as long as it is positive. This test is shown in Figure 2, with a both a constant  $K(z)$ , and a smooth step function between two values of  $K$ . For both, the constant initial concentration is close to unchanged, save for numerical errors.

The systems should also, given  $k_w = 0$ , conserve mass. To test this, a initial distribution of two Gaussian functions were evolved in time, with a non-constant diffusivity. The result is shown in Figure 3, where the mass is conserved to a good approximation.

A sharply peaked Gaussian package should have a variance that increases linearly with time, then approach a steady state. This is shown in Figure 4.

Figure 5 shows the depletion of  $\text{CO}_2$  from the ocean, given a zero partial pressure in the atmosphere. For a small Biot number, this should follow a exponential decay, as this test shows.

Lastly, Figure 6 shows how the ocean reaches a equilibrium with the atmosphere, given a non-zero, positive mass-transfer coefficient  $k_w$  and either constant or non-constant diffusivity.

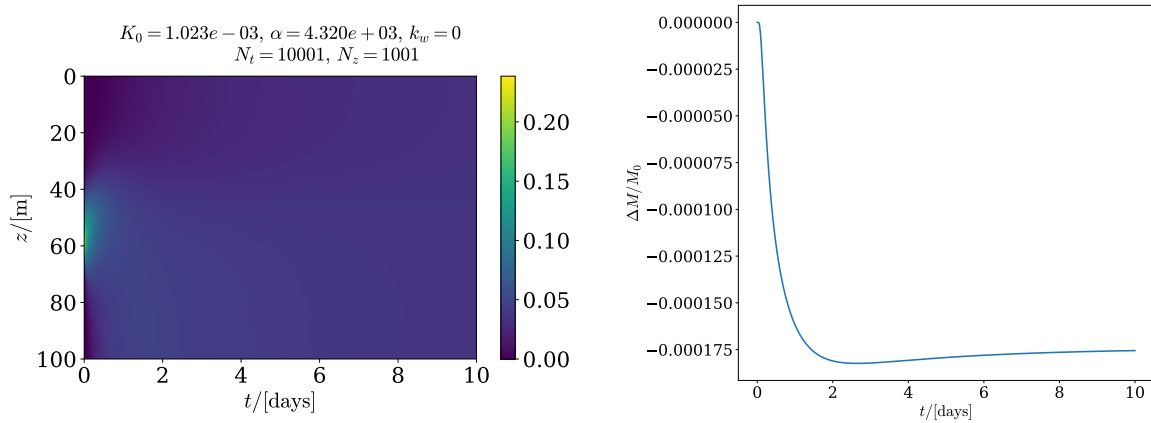


Figure 3: The evolution of a Gaussian distribution of CO<sub>2</sub> is shown on the left. On the right, the relative change in mass as a function of time is plotted.

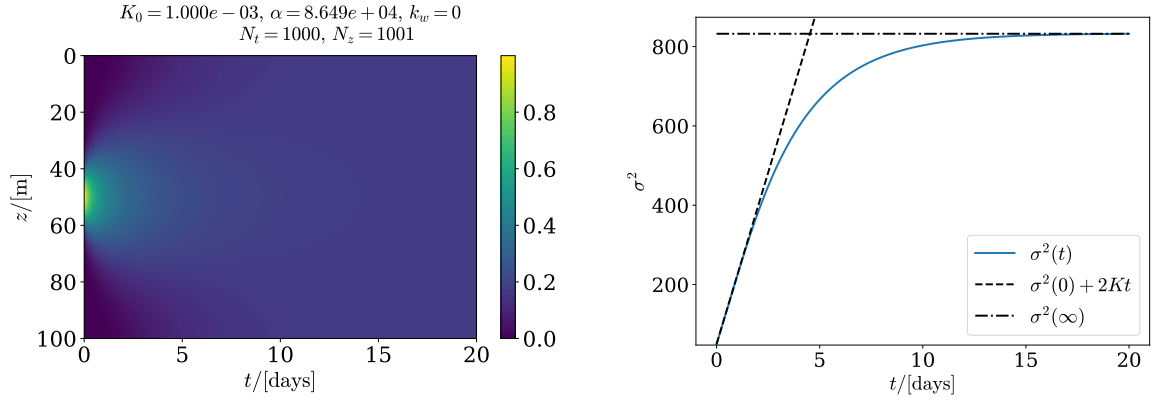


Figure 4: A system with an initial Gaussian distribution is simulated. On the right, the variance as a function of time is shown. The increase is initially constant, but then approaches a steady state.

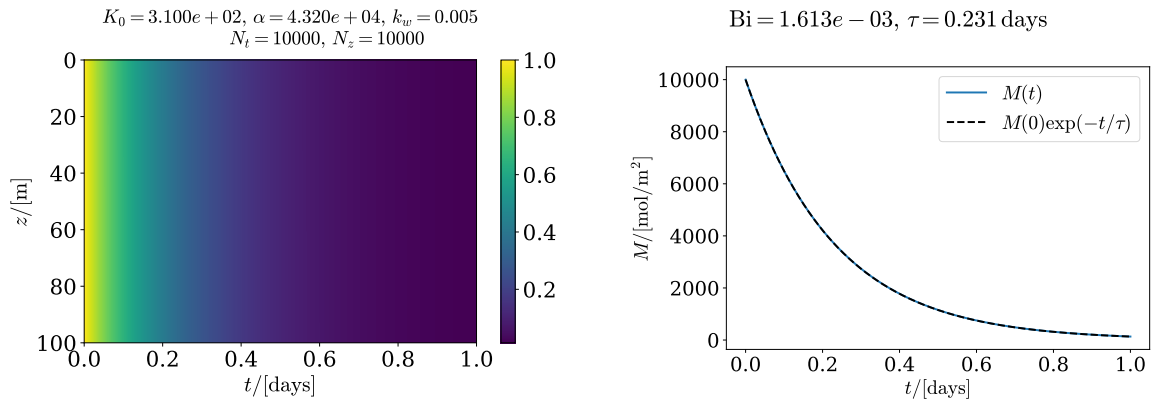


Figure 5: The slow removal of CO<sub>2</sub> from the ocean, when the atmosphere contains a partial pressure of 0.

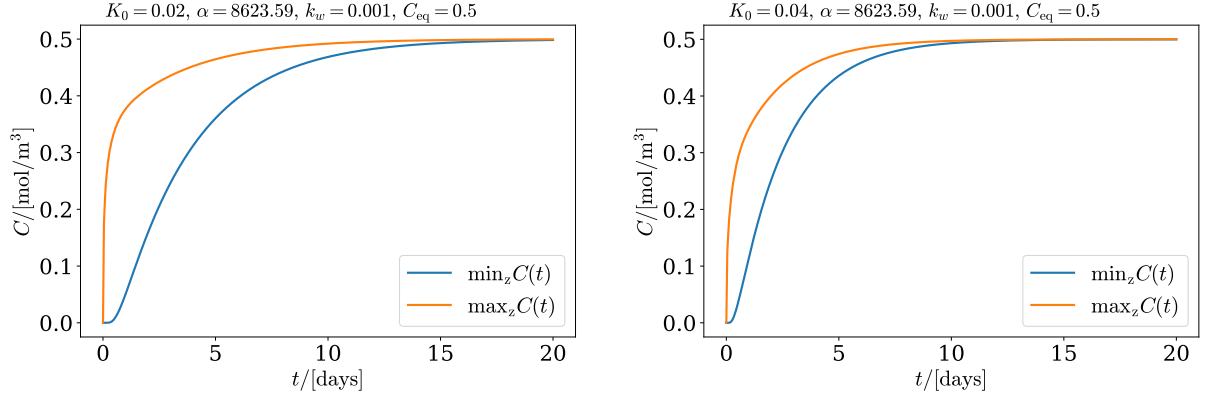


Figure 6: Equilibration of the ocean with a atmosphere with  $C_{\text{eq}} = 0.5$ . The figure on the left is a system with constant diffusivity, the left side has a oscillating but always positive diffusivity.

## Results

### Shallow waters

To simulate the uptake of  $\text{CO}_2$  in shallow areas,  $L = 100\text{ m}$ , diffusivity  $K(z)$  is used as given in equation (11) in [1]. Figure 7 illustrates the convergence of the simulation, which shows that the simulation still has quadratic convergence both in time and space. The convergence test are run for a time of 10 days.

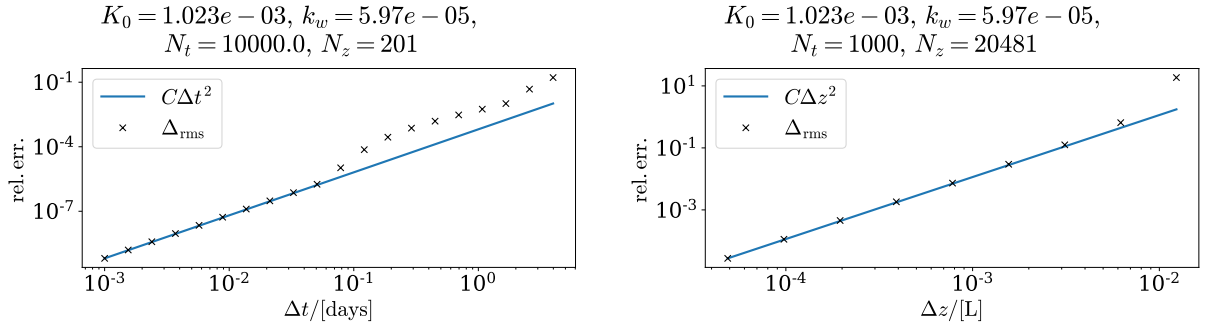


Figure 7: Error, as a function of step length in both time and space.

The simulation is then run for 180 days, with  $N_t = N_z = 10000$ . This corresponds to  $\Delta t = 180/10000 = 1.8 \cdot 10^{-2}\text{ days}$ , and  $\Delta z = L/10000 = 10^{-4}L$ . The results are shown in Figure 8, together with the form of the diffusivity,  $K(z)$ . Figure 9 shows how the minimum and maximum concentration evolves as a function of time, together with some snap shots of the concentration as it evolves. The surface of the ocean, where wind dominates the diffusion, first absorbs the  $\text{CO}_2$ . The deeper parts of the ocean lacks behind some, as there low diffusivity around 40 meters are a bottle neck, but it does not take long before it penetrates down to the ocean floor.

At 90 days, the ocean is nearly in equilibrium with the atmosphere, and at 180 it is so close as does not make a difference. This gives confidence to an assumption that the carbon in the ocean is in sync with the atmosphere.

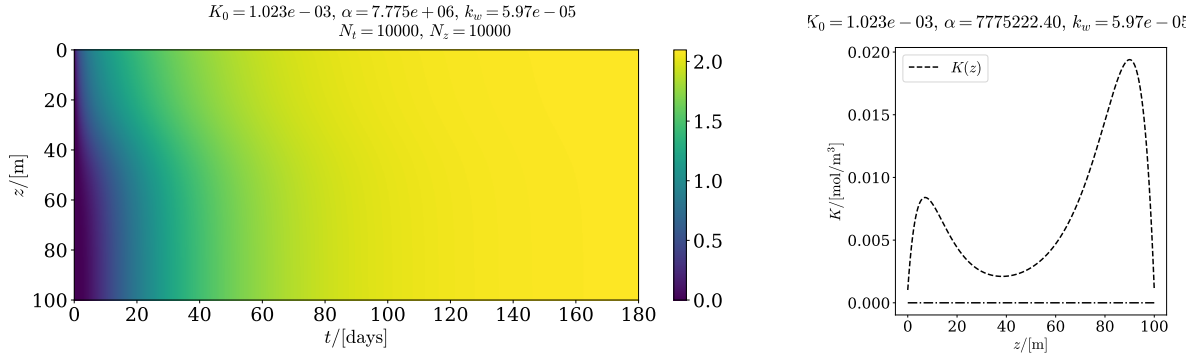


Figure 8: cap

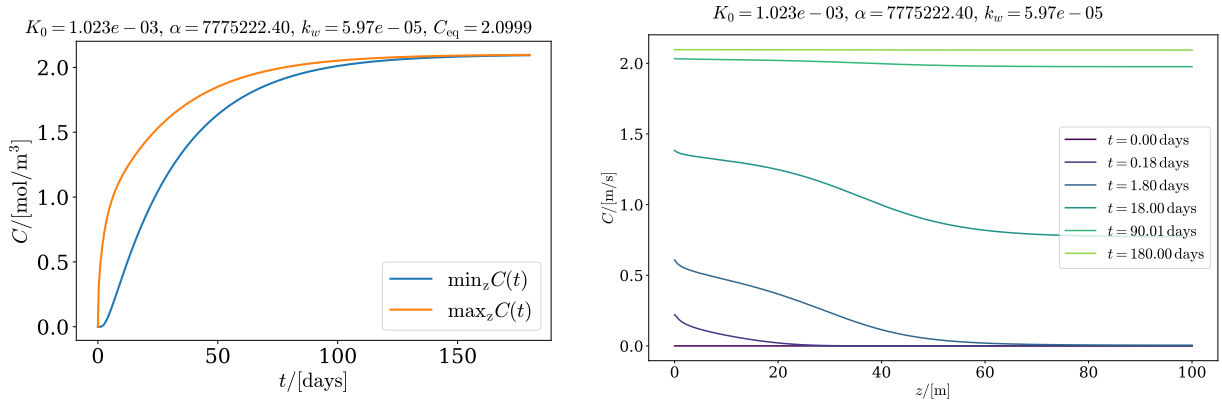


Figure 9: cap

Deep waters

## References

- [1] Exercise 3, tfy4235 computational physics, 2021.