## Climate Modeling

October 24, 2024

### PRACTICAL REPORT

#### MODEL HIERARCHY & SIMPLIFIED CLIMATE MODELS

By:

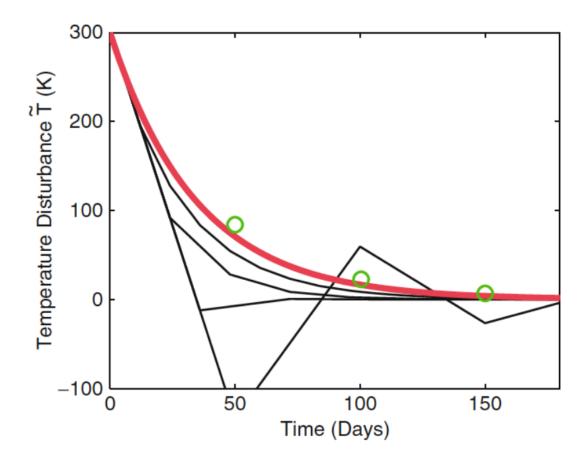
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### Contents

1	Plot the analytical red curve	3
2	Overlay the numerical solutions using the Euler scheme with different time steps of 12, 24, 36, 50 days	4
3	Overlay the results from the Runge-Kutta scheme with $\Delta t = 50$ days	5
4	Conclusion	8

Reproduce the following figure.



Numerical solutions of (7) with initial perturbation = 300K computed with the Euler scheme and time steps of 12, 24, 36, 50 days. The analytical solution is in red; the results from the classical Runge-Kutta scheme ( t = 50days) are labelled with green circles.

Initial perturbation 300K, tau=35 days

#### 1 Plot the analytical red curve

$$T = ae^{-\frac{t}{\tau}}$$

where:

$$\tau = 35$$
$$a = T(0) = 300k$$

For analytical result, I choose dt = 50

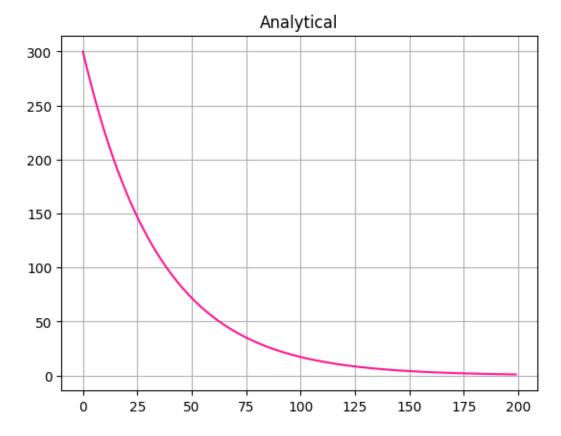
[1]: import numpy as np import matplotlib.pyplot as plt

```
dt_rk = 50

t_ana = np.arange(0, 200)

Ts_ana = T0 * np.exp(-t_ana / tau)

plt.plot(t_ana, Ts_ana, color='deeppink')
plt.title('Analytical')
plt.grid()
plt.show()
```



# 2 Overlay the numerical solutions using the Euler scheme with different time steps of 12, 24, 36, 50 days

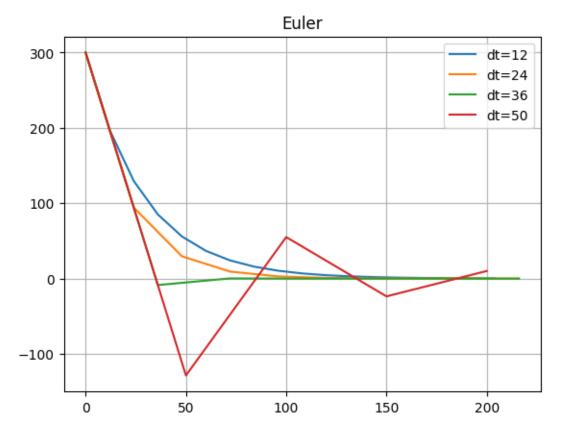
$$T_{x+1} = T_x - \frac{T_x \cdot dt}{\tau}$$

I repeat this calculation for different time step 12, 24, 36, and 50.

```
Ts = np.zeros(len(t))
Ts[0] = T0

for i in range(1, len(t)):
    Ts[i] = Ts[i - 1] * (1 - dt / tau)
    euler.append((t, Ts, dt))
    plt.plot(t, Ts, label=f'dt={dt}')

plt.title('Euler')
plt.grid()
plt.legend()
plt.show()
```



Lower dt gives much better approximation than high dt.

# 3 Overlay the results from the Runge-Kutta scheme with $\Delta t = 50$ days

Differential equation:

$$f(x,y) = -\frac{1}{\tau}y$$

Runge Kutta:

$$F(x_n, y_n) = \frac{1}{6} \left( K_1 + 2K_2 + 2K_3 + K_4 \right)$$

where:

$$\begin{split} K_1 &= f(x_n,y_n) \\ K_2 &= f\left(x_n + \frac{1}{2}\Delta x, y_n + \frac{1}{2}\Delta x K_1\right) \\ K_3 &= f\left(x_n + \frac{1}{2}\Delta x, y_n + \frac{1}{2}\Delta x K_2\right) \\ K_4 &= f\left(x_n + \Delta x, y_n + \Delta x K_3\right) \end{split}$$

In the code it creates a function that takes three inputs: time, temperature, and time step. It use a loop to calculate the temperature at each point using the Runge-Kutta formula:

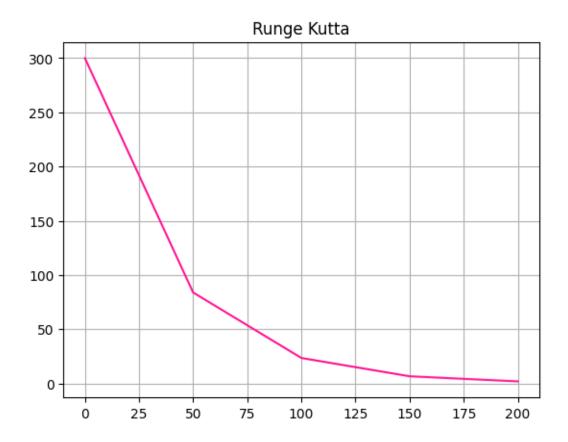
$$T_{x+1} = T_x + dt \cdot F(t_i, T_i, dt)$$

```
[5]: def runge_kutta_step(t, T, dt):
    K1 = f(t, T)
    K2 = f(t + 0.5 * dt, T + 0.5 * dt * K1)
    K3 = f(t + 0.5 * dt, T + 0.5 * dt * K2)
    K4 = f(t + dt, T + dt * K3)
    return (K1 + 2 * K2 + 2 * K3 + K4) / 6

t_rk = np.arange(0, 200 + dt_rk, dt_rk)
T_rk = np.zeros(len(t_rk))
T_rk[0] = T0

for i in range(len(t_rk) - 1):
    T_rk[i + 1] = T_rk[i] + dt_rk * runge_kutta_step(t_rk[i], T_rk[i], dt_rk)
plt.plot(t_rk, T_rk, color='deeppink')

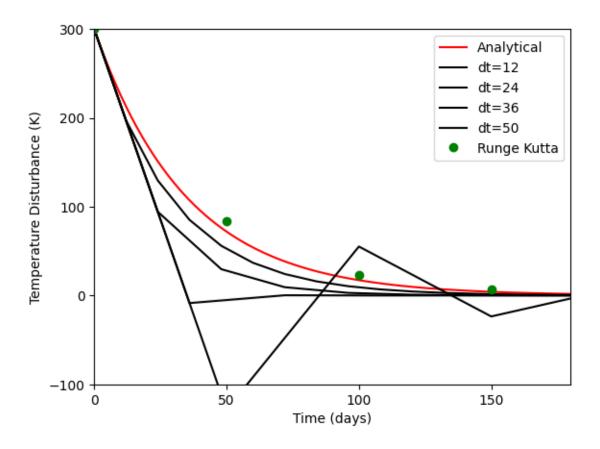
plt.title('Runge Kutta')
plt.grid()
plt.show()
```



Compare to the Euler method curve with the same dt=50, Runge Kutta method is more accurate. Then I plot all functions and try to make it looks similar as the given figure as much as possible.

```
[6]: plt.plot(t_ana, Ts_ana, label="Analytical", color='red')
    for t, Ts, dt in euler:
        plt.plot(t, Ts, label=f'dt={dt}', color='black')

plt.plot(t_rk, T_rk, 'o', label='Runge Kutta', color='green')
    plt.xlim(0, 180)
    plt.xticks([0, 50, 100, 150])
    plt.ylim(-100, 300)
    plt.yticks([-100, 0, 100, 200, 300])
    plt.xlabel("Time (days)")
    plt.ylabel("Temperature Disturbance (K)")
    plt.legend()
    plt.show()
```



#### 4 Conclusion

- For the Euler method, if dt decreases it gives a curve that looks the most like the analytical one, which means a better approximation.
- The Runge Kutta method is much better than the Euler method, with the same dt=50, for Euler it doesn't even have the right shape, but for Runge Kutta the overall shape follows the analytical one really close.