AP7

Loewner differential equation

The Loewner differential equation is written as:

$$rac{\partial g_t(z)}{\partial t} = rac{2}{g_t(z) - \lambda(t)},$$

where:

- $g_t(z)$ is the Loewner chain,
- $\lambda(t)$ is the driving function (often continuous or piecewise).

For numerical purposes, this equation can be reformulated into a system of ODEs. Let's assume a toy driving function like $\lambda(t)=2t$ to simplify implementation.

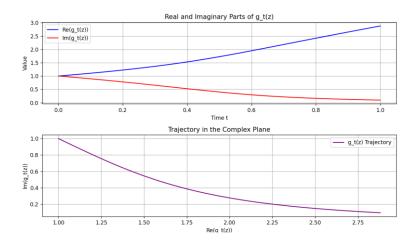
Rewriting:

$$g_t(z) = y(t), \quad ext{so the system becomes:} \quad rac{dy}{dt} = rac{2}{y - \lambda(t)}.$$

https://en.wikipedia.org/wiki/List_of_nonlinear_ordinary_differential_equations

Result:

The numerical solution and visualization of the Loewner differential equation demonstrate the impact of the driving function $\lambda(t)=2t \cdot (t)=2t$ on the evolution of $gt(z)g_t(z)g_t(z)$. The results highlight the structured yet dynamic nature of the Loewner chains, with the driving function dictating the behavior of the solution in both the realimaginary space and the complex plane. These visualizations provide insights into how Loewner's theory models the growth of conformal maps over time.



Code:

```
import numpy as np
import matplotlib.pyplot as plt
# Loewner differential equation
def loewner_rhs(y, t, lambda_t):
    return 2 / (y - lambda_t)
# driving function
def driving_function(t):
    return 2 * t
# Adams-Bashforth 2-step method implementation
def adams_bashforth_2step(f, y0, t0, t_end, h, lambda_func):
   t = np.arange(t0, t_end + h, h)
   n_steps = len(t)
   y = np.zeros(n_steps, dtype=complex)
   y[0] = y0
    lambda_0 = lambda_func(t[0])
    y[1] = y[0] + h * f(y[0], t[0], lambda_0)
    for n in range(1, n_steps - 1):
        lambda n = lambda func(t[n])
        lambda_nm1 = lambda_func(t[n - 1])
        y[n + 1] = y[n] + h * (
            (3 / 2) * f(y[n], t[n], lambda_n) - (1 / 2) * f(y[n - 1], t[n - 1],
lambda nm1)
    return t, y
# Parameters for the toy model
initial_condition = 1 + 1j
t0, t_end = 0, 1
h = 0.01
# call Loewner differential equation
```

```
t, y = adams_bashforth_2step(loewner_rhs, initial_condition, t0, t_end, h,
driving_function)
# Visualization
plt.figure(figsize=(10, 6))
# Plot the real and imaginary parts
plt.subplot(2, 1, 1)
plt.plot(t, y.real, label="Re(g_t(z))", color="blue")
plt.plot(t, y.imag, label="Im(g_t(z))", color="red")
plt.xlabel("Time t")
plt.ylabel("Value")
plt.title("Real and Imaginary Parts of g_t(z)")
plt.legend()
plt.grid()
# Plot the trajectory in the complex plane
plt.subplot(2, 1, \overline{2})
plt.plot(y.real, y.imag, label="g_t(z) Trajectory", color="purple")
plt.xlabel("Re(g_t(z))")
plt.ylabel("Im(g_t(z))")
plt.title("Trajectory in the Complex Plane")
plt.legend()
plt.grid()
plt.tight_layout()
plt.show()
```