

Programming of Differential Equations (Appendix E)

Hans Petter Langtangen

Simula Research Laboratory
University of Oslo, Dept. of Informatics

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Differential equations

- A differential equation (ODE) written in generic form:

$$u'(t) = f(u(t), t)$$

The solution of this equation is a function $u(t)$

- To obtain a unique solution $u(t)$, the ODE must have an initial condition: $u(0) = U_0$
- Different choices of $f(u, t)$ give different ODEs:

$$f(u, t) = \alpha u, \quad u' = \alpha u \quad \text{exponential growth}$$

$$f(u, t) = \alpha u \left(1 - \frac{u}{R}\right), \quad u' = \alpha u \left(1 - \frac{u}{R}\right) \quad \text{logistic growth}$$

$$f(u, t) = -b|u|u + g, \quad u' = -b|u|u + g \quad \text{body in fluid}$$

Our task: solve any ODE $u' = f(u, t)$ by programming

How to solve a general ODE numerically

- Given $u' = f(u, t)$ and $u(0) = U_0$, the Forward Euler method generates a sequence of u_1, u_2, \dots, u_n values for u at times t_1, t_2, \dots, t_n :

$$u_{k+1} = u_k + \Delta t f(u_k, t_k)$$

where $t_k = k\Delta t$

- This is a simple stepping-forward-in-time formula
- Algorithm using growing lists for u_k and t_k :
 - Create arrays u and t to hold u_k and t_k for $k = 0, 1, 2, \dots, n$
 - Set initial condition: $u[0] = U_0$
 - For $k = 0, 1, 2, \dots, n-1$:
 - $u[k+1] = u[k] + dt * f(u[k], t[k])$
 - $t[k+1] = t[k] + dt$

Implementation as a function

```
def ForwardEuler(f, U0, T, n):  
    """Solve u'=f(u,t), u(0)=U0, with n steps until t=T."""  
    import numpy as np  
    t = np.zeros(n+1)  
    u = np.zeros(n+1) # u[k] is the solution at time t[k]  
    u[0] = U0  
    t[0] = 0  
    dt = T/float(n)  
    for k in range(n):  
        t[k+1] = t[k] + dt  
        u[k+1] = u[k] + dt*f(u[k], t[k])  
    return u, t
```

This simple function can solve any ODE (!)

Example on using the function

Mathematical problem:

Solve $u' = u$, $u(0) = 1$, for $t \in [0, 3]$, with $\Delta t = 0.1$

Basic code:

```
def f(u, t):  
    return u  
  
U0 = 1  
T = 3  
n = 30  
u, t = ForwardEuler(f, dt, U0, T)
```

Compare exact and numerical solution:

```
from scitools.std import plot, exp  
u_exact = exp(t)  
plot(t, u, 'r-', t, u_exact, 'b-',  
      xlabel='t', ylabel='u', legend=('numerical', 'exact'),  
      title="Solution of the ODE u'=u, u(0)=1")
```

A class for solving ODEs

- Instead of a function for solving any ODE we now want to make a class
- Usage of the class:

```
method = ForwardEuler(f, dt)  
method.set_initial_condition(U0, t0)  
u, t = method.solve(T)
```
- Store f , Δt , and the sequences u_k , t_k as attributes
- Split the steps in the `ForwardEuler` function into three methods

The code for a class for solving ODEs (part 1)

```
import numpy as np

class ForwardEuler_v1:
    def __init__(self, f, U0, T, n):
        self.f, self.U0, self.T, self.n = f, dt, U0, T, n
        self.dt = T/float(n)
        self.u = np.zeros(n+1)
        self.t = np.zeros(n+1)
```

The code for a class for solving ODEs (part 2)

```
class ForwardEuler_v1:
    ...
    def solve(self):
        """Compute solution for 0 <= t <= T."""
        self.u[0] = float(self.U0)
        self.t[0] = float(0)

        for k in range(self.n):
            self.k = k
            self.t[k+1] = self.t[k] + self.dt
            self.u[k+1] = self.advance()
        return self.u, self.t

    def advance(self):
        """Advance the solution one time step."""
        u, dt, f, k, t = self.u, self.dt, self.f, self.k, self.t

        unew = u[k] + dt*f(u[k], t[k])
        return unew
```

Alternative class code solving ODEs (part 1)

```
import numpy as np

class ForwardEuler:
    def __init__(self, f):
        if not callable(f):
            raise TypeError('f is %s, not a function' % type(f))
        self.f = f

    def set_initial_condition(self, U0):
        self.U0 = float(U0)
```

Alternative class code for solving ODEs (part 2)

```
class ForwardEuler:
    ...
    def solve(self, time_points):
        """Compute u for t values in time_points list."""
        self.t = np.asarray(time_points)
        self.u = np.zeros(len(time_points))
        # Assume time_points[0] corresponds to self.U0
        self.u[0] = self.U0

        for k in range(len(self.t)-1):
            self.k = k
            self.u[k+1] = self.advance()
        return self.u, self.t

    def advance(self):
        """Advance the solution one time step."""
        u, f, k, t = self.u, self.f, self.k, self.t
        dt = self.t[k+1] - t[k]
        unew = u[k] + dt*f(u[k], t[k])
        return unew
```

Verifying the class implementation

Mathematical problem:

$$u' = 0.2 + (u - h(t))^4, \quad u(0) = 3, \quad t \in [0, 3]$$

$$u(t) = h(t) = 0.2t + 3 \quad (\text{exact solution})$$

The Forward Euler method will reproduce such a linear u exactly!

Code:

```
def f(u, t):
    return 0.2 + (u - h(t))**4

def h(t):
    return 0.2*t + 3

U0 = 3; dt = 0.4; T = 3; n = int(round(T/dt))
method = ForwardEuler(f)
method.set_initial_condition(U0)
time_points = np.linspace(0, T, n)
u, t = method.solve(time_points)
u_exact = h(t)
print 'Numerical: %s\nExact: %s' % (u, u_exact)
```

Using a class to hold the right-hand side $f(u, t)$

Mathematical problem:

$$u'(t) = \alpha u(t) \left(1 - \frac{u(t)}{R}\right), \quad u(0) = U_0, \quad t \in [0, 40]$$

Class for right-hand side $f(u, t)$:

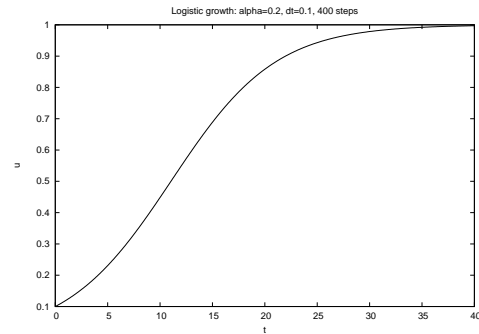
```
class Logistic:
    def __init__(self, alpha, R, U0):
        self.alpha, self.R, self.U0 = alpha, float(R), U0

    def __call__(self, u, t): # f(u,t)
        return self.alpha*u*(1 - u/self.R)
```

Main program:

```
problem = Logistic(0.2, 1, 0.1)
time_points = np.linspace(0, 40, 401)
method = ForwardEuler(problem)
method.set_initial_condition(problem.U0)
u, t = method.solve(time_points)
```

Figure of the solution



Ordinary differential equations

Mathematical problem:

$$u'(t) = f(u, t)$$

Initial condition:

$$u(0) = U_0$$

Possible applications:

- Exponential growth of money or populations: $f(u, t) = \alpha u$, $\alpha = \text{const}$
- Logistic growth of a population under limited resources:

$$f(u, t) = \alpha u \left(1 - \frac{u}{R}\right)$$

where R is the maximum possible value of u

- Radioactive decay of a substance: $f(u, t) = -\alpha u$, $\alpha = \text{const}$

Numerical solution of ordinary differential equations

- Numerous methods for $u'(t) = f(u, t)$, $u(0) = U_0$
- The Forward Euler method:

$$u_{k+1} = u_k + \Delta t f(u_k, t_k)$$

- The 4th-order Runge-Kutta method:

$$u_{k+1} = u_k + \frac{1}{6} (K_1 + 2K_2 + 2K_3 + K_4)$$

$$K_1 = \Delta t f(u_k, t_k),$$

$$K_2 = \Delta t f\left(u_k + \frac{1}{2}K_1, t_k + \frac{1}{2}\Delta t\right),$$

$$K_3 = \Delta t f\left(u_k + \frac{1}{2}K_2, t_k + \frac{1}{2}\Delta t\right),$$

$$K_4 = \Delta t f(u_k + K_3, t_k + \Delta t)$$

- There is a jungle of different methods – how to program?

A superclass for ODE methods

Common tasks for ODE solvers:

- Store the solution u_k and the corresponding time levels t_k , $k = 0, 1, 2, \dots, N$
- Store the right-hand side function $f(u, t)$
- Store the time step Δt and last time step number k
- Set the initial condition
- Implement the loop over all time steps

- Code for the steps above are common to all classes and hence placed in superclass `ODESolver`
- Subclasses, e.g., `ForwardEuler`, just implement the specific stepping formula in a method `advance`

The superclass code

```
class ODESolver:
    def __init__(self, f):
        self.f = f

    def advance(self):
        """Advance solution one time step."""
        raise NotImplementedError # implement in subclass

    def set_initial_condition(self, U0):
        self.U0 = float(U0)

    def solve(self, time_points):
        self.t = np.asarray(time_points)
        self.u = np.zeros(len(self.t))
        # Assume that self.t[0] corresponds to self.U0
        self.u[0] = self.U0

        # Time loop
        for k in range(n-1):
            self.k = k
            self.u[k+1] = self.advance()
        return self.u, self.t
```

Implementation of the Forward Euler method

Subclass code:

```
class ForwardEuler(ODESolver):
    def advance(self):
        u, dt, f, k, t = self.u, self.dt, self.f, self.k, self.t
        unew = u[k] + dt*f(u[k], t)
        return unew
```

Application code for $u' - u = 0$, $u(0) = 1$, $t \in [0, 3]$, $\Delta t = 0.1$:

```
from ODESolver import ForwardEuler
def test1(u, t):
    return u

method = ForwardEuler(test1)
method.set_initial_condition(U0=1)
u, t = method.solve(time_points=linspace(0, 3, 31))
plot(t, u)
```

The implementation of a Runge-Kutta method

Subclass code:

```
class RungeKutta4(ODESolver):
    def advance(self):
        u, dt, f, k, t = self.u, self.dt, self.f, self.k
        dt2 = dt/2.0
        K1 = dt*f(u[k], t)
        K2 = dt*f(u[k] + 0.5*K1, t + dt2)
        K3 = dt*f(u[k] + 0.5*K2, t + dt2)
        K4 = dt*f(u[k] + K3, t + dt)
        unew = u[k] + (1/6.0)*(K1 + 2*K2 + 2*K3 + K4)
        return unew
```

Application code (same as for ForwardEuler):

```
from ODESolver import RungeKutta4
def test1(u, t):
    return u

method = RungeKutta4(test1)
method.set_initial_condition(U0=1)
u, t = method.solve(time_points=linspace(0, 3, 31))
plot(t, u)
```

Including user-defined termination of the solution process

- Sometimes a property of the solution determines when to stop the solution process (e.g., when $u < 10^{-7} \approx 0$)
- Extension solve(time_points, terminate)

```
def terminate(u, t, step_no):
    tol = 1.0E-6 # small number
    if abs(u[step_no]) < tol: # close enough to zero?
        return True
    else:
        return False
```

- Break time loop if terminate(u, t, k) is true

Making a flexible toolbox for solving ODEs

- We can continue to implement formulas for different numerical methods for ODEs – a new method just requires the formula, not the rest of the code needed to set initial conditions and loop in time
- The OO approach saves typing – no code duplication
- Challenge: you need to understand exactly which "slots" in subclasses you have to fill in – the overall code is an interplay of the superclass and the subclass
- Warning: more sophisticated methods for ODEs do not fit straight into our simple superclass – a more sophisticated superclass is needed, but the basic ideas of using OO remain the same
- Believe our conclusion: ODE methods are best implemented in a class hierarchy!

Example on a system of ODEs

- Several coupled ODEs make up a *system of ODEs*
- A simple example:

$$\begin{aligned}u'(t) &= v(t), \\v'(t) &= -u(t)\end{aligned}$$

Two ODEs with two unknowns $u(t)$ and $v(t)$

- Each unknown must have an initial condition, say

$$u(0) = 0, \quad v(0) = 1$$

- One can then derive the exact solution

$$u(t) = \sin(t), \quad v(t) = \cos(t)$$

- Systems of ODEs appear frequently in physics, biology, finance, ...

Another example on a system of ODEs

- Second-order ordinary differential equation, for a spring-mass system,

$$mu'' + \beta u' + ku = 0, \quad u(0) = U_0, \quad u'(0) = 0$$

- We can rewrite this as a system of two *first-order* equations
- Introduce two new unknowns

$$u^{(0)}(t) \equiv u(t), \quad u^{(1)}(t) \equiv u'(t)$$

- The first-order system is then

$$\begin{aligned}\frac{d}{dt}u^{(0)}(t) &= u^{(1)}(t), \\ \frac{d}{dt}u^{(1)}(t) &= -m^{-1}\beta u^{(1)} - m^{-1}ku^{(0)}\end{aligned}$$

$$u^{(0)}(0) = U_0, \quad u^{(1)}(0) = 0$$

Vector notation for systems of ODEs (part 1)

- In general we have n unknowns

$$u^{(0)}(t), u^{(1)}(t), \dots, u^{(n-1)}(t)$$

in a system of n ODEs:

$$\begin{aligned}\frac{d}{dt}u^{(0)} &= f^{(0)}(u^{(0)}, u^{(1)}, \dots, u^{(n-1)}, t) \\ \frac{d}{dt}u^{(1)} &= f^{(1)}(u^{(0)}, u^{(1)}, \dots, u^{(n-1)}, t) \\ &\dots = \dots \\ \frac{d}{dt}u^{(n-1)} &= f^{(n-1)}(u^{(0)}, u^{(1)}, \dots, u^{(n-1)}, t)\end{aligned}$$

Vector notation for systems of ODEs (part 2)

- We can collect the $u^{(i)}(t)$ functions and right-hand side functions $f^{(i)}$ in vectors:

$$u = (u^{(0)}, u^{(1)}, \dots, u^{(n-1)})$$

$$f = (f^{(0)}, f^{(1)}, \dots, f^{(n-1)})$$

- The first-order system can then be written

$$u' = f(u, t), \quad u(0) = U_0$$

where u and f are vectors and U_0 is a vector of initial conditions

- Why is this notation useful? The notation make a scalar ODE and a system look the same, and we can easily make Python code that can handle both cases within the same lines of code (!)

How to make class ODESolver work for systems

- Recall: ODESolver was written for a scalar ODE
- Now we want it to work for a system $u' = f$, $u(0) = U_0$, where u , f and U_0 are vectors (arrays)
- Forward Euler for a system:

$$u_{k+1} = u_k + \Delta t f(u_k, t_k)$$

(vector = vector + scalar × vector)

- In Python code:
`u_new = u[k] + dt*f(u[k], t)`
 where u is a list of arrays ($u[k]$ is an array) and f is a function returning an array (all the right-hand sides $f^{(0)}, \dots, f^{(n-1)}$)
- Result: ODESolver will work for systems!
- The only change: ensure that $f(u, t)$ returns an array (This can be done by a general adjustment in the superclass!)

The superclass code (part 1)

```
class ODESolver:
    def __init__(self, f):
        self.f = lambda u, t: np.asarray(f(u, t), float)

    def advance(self):
        """Advance solution one time step."""
        raise NotImplementedError # implement in subclass

    def set_initial_condition(self, U0):
        if isinstance(U0, (float,int)): # scalar ODE
            self.n = 1
            U0 = float(U0)
        else: # system of ODEs
            U0 = np.asarray(U0)
            self.n = U0.size
            self.U0 = U0
```

The superclass code (part 2)

```
class ODESolver:
    ...
    def solve(self, time_points, terminate=None):
        if terminate is None:
            terminate = lambda u, t, step_no: False

        self.t = np.asarray(time_points)
        n = self.t.size
        if self.n == 1: # scalar ODEs
            self.u = np.zeros(n)
        else: # systems of ODEs
            self.u = np.zeros((n,self.n))

        # Assume that self.t[0] corresponds to self.U0
        self.u[0] = self.U0

        # Time loop
        for k in range(n-1):
            self.k = k
            self.u[k+1] = self.advance()
            if terminate(self.u, self.t, self.k+1):
                break # terminate loop over k
        return self.u, self.t
```

Back to implementing a system (part 1)

Spring-mass system formulated as a system of ODEs:

$$m u'' + \beta u' + k u = 0, \quad u(0), u'(0) \text{ known}$$

$$u^{(0)} = u, \quad u^{(1)} = u'$$

$$u(t) = (u^{(0)}(t), u^{(1)}(t))$$

$$f(u, t) = (u^{(1)}(t), -m^{-1}\beta u^{(1)} - m^{-1}k u^{(0)})$$

$$u'(t) = f(u, t)$$

Code defining the right-hand side:

```
def myf(u, t):
    # u is array with two components u[0] and u[1]:
    return [u[1],
            -beta*u[1]/m - k*u[0]/m]
```

Back to implementing a system (part 2)

Better (no global variables):

```
class MyF:
    def __init__(self, m, k, beta):
        self.m, self.k, self.beta = m, k, beta

    def __call__(self, u, t):
        m, k, beta = self.m, self.k, self.beta
        return [u[1], -beta*u[1]/m - k*u[0]/m]
```

Main program:

```
from ODESolver import ForwardEuler
# initial condition:
U0 = [1.0, 0]
f = MyF(1.0, 1.0, 0.0) # u'' + u = 0 => u(t)=cos(t)
T = 4*pi; dt = pi/20; n = int(round(T/dt))
method = ForwardEuler(f)
method.set_initial_condition(U0)
u, t = method.solve(time_points=linspace(0, T, n))

# u is an array of [u0,u1] arrays, plot all u0 values:
u0_values = u[:,0]
u0_exact = cos(t)
plot(t, u0_values, 'r-', t, u0_exact, 'b-')
```

Application: throwing a ball (part 1)

- Newton's 2nd law for a ball's trajectory through air leads to

$$\begin{aligned}\frac{dx}{dt} &= v_x \\ \frac{dv_x}{dt} &= 0 \\ \frac{dy}{dt} &= v_y \\ \frac{dv_y}{dt} &= -g\end{aligned}$$

Air resistance is neglected but can easily be added!

- 4 ODEs with 4 unknowns: the ball's position $x(t)$, $y(t)$ and the velocity $v_x(t)$, $v_y(t)$

Application: throwing a ball (part 2)

Define the right-hand side:

```
def f(u, t):  
    x, vx, y, vy = u  
    g = 9.81  
    return [vx, 0, vy, -g]
```

Main program:

```
from ODESolver import ForwardEuler  
# t=0: prescribe velocity magnitude and angle  
v0 = 5; theta = 80*pi/180  
# initial condition:  
u0 = [0, v0*cos(theta), 0, v0*sin(theta)]  
  
T = 1.2; dt = 0.01  
method = ForwardEuler(f, dt)  
method.set_initial_condition(u0)  
u, t = method.solve(T)  
# u is an array of [x,vx,y,vy] arrays, plot y vs x:  
plot(u[:,0], u[:,1])
```

Application: throwing a ball (part 3)

Comparison of exact and Forward Euler solutions

