

Scientific Computing for Differential Equations

Lecture 02C - The Implicit Euler Method

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Ordinary Differential Equations

Explicit Euler method with fixed step size

The initial value problem (IVP)

$$x(t_a) = x_a \quad (1a)$$

$$\dot{x}(t) = f(t, x(t)) \quad t_a \leq t \leq t_b \quad (1b)$$

can be solved using Euler's explicit method

$$t_0 = t_a, \quad x_0 = x_a \quad (2a)$$

$$t_{k+1} = t_k + \Delta t, \quad x_{k+1} = x_k + \Delta t f(t_k, x_k) \quad (2b)$$

with a fixed time-step, Δt , using the Matlab code

```
1 function [T,X] = ExplicitEulerFixedStepSize(fun,ta,tb,N,xa,varargin)
2
3 % Compute step size and allocate memory
4 dt = (tb-ta)/N;
5 nx = size(xa,1);
6 X = zeros(nx,N+1);
7 T = zeros(1,N+1);
8
9 % Eulers Explicit Method
10 T(:,1) = ta;
11 X(:,1) = xa;
12 for k=1:N
13     f = feval(fun,T(k),X(:,k),varargin{:});
14     T(:,k+1) = T(:,k) + dt;
15     X(:,k+1) = X(:,k) + f*dt;
16 end
17
18 % Form a nice table for the result
19 T = T';
20 X = X';
```

Newton's Method

We want to find a root of the multivariate function

$$R(x) = 0 \quad R : \mathbb{R}^n \mapsto \mathbb{R}^n \quad (3)$$

We make a first order Taylor expansion of R around $x^{[k]}$

$$R(x) \approx R(x^{[k]}) + \frac{\partial R}{\partial x}(x^{[k]})(x - x^{[k]}) = 0 \quad (4)$$

and set that first order Taylor approximation to zero.

Newton's method is

$$R(x^{[k+1]}) \approx R(x^{[k]}) + \frac{\partial R}{\partial x}(x^{[k]}) \overbrace{(x^{[k+1]} - x^{[k]})}^{\Delta x} = 0 \quad (5a)$$

$$x^{[k+1]} = x^{[k]} + \Delta x \quad (5b)$$

that can be expressed as

$$b := R(x^{[k]}) \quad A := \frac{\partial R}{\partial x}(x^{[k]}) \quad (6a)$$

$$\text{Solve for } \Delta x: \quad A\Delta x = b \quad (6b)$$

$$x^{[k+1]} := x^{[k]} - \Delta x \quad (6c)$$

Numerical Solution of Linear Systems of Equations

$$Ax = b \quad (7)$$

- ▶ Backslash in Matlab: $x = A \backslash b$
- ▶ LU decomposition

$$PA = LU \quad (8a)$$

$$PAx = L \overbrace{Ux}^{=y} = Pb \quad (8b)$$

1. Compute $\bar{b} = Pb$
2. Solve for y : $Ly = \bar{b}$
3. Solve for x : $Ux = y$

Matlab:

```
[L,U,p] = lu(A,'vector');    % factorization  
x = L \ (U \ b(p));          % back-substitution
```

Newton's Method - Matlab

The solution of

$$R(x) = 0 \quad (9)$$

can be accomplished by the Matlab code

```
1      function x = NewtonsMethod(ResidualFunJac, x0, tol, maxit, varargin)
2
3      k = 0;
4      x = x0;
5      [R,dRdx] = feval(ResidualFunJac,x,varargin{:})
6      while( (k < maxit) & (norm(R,'inf') > tol) )
7          k = k+1;
8          dx = dRdx\R;
9          x = x - dx;
10         [R,dRdx] = feval(ResidualFunJac,x,varargin{:});
11     end
```

This code terminates if $\|R(x)\|_{\infty} \leq \text{tol}$ or $k \geq \text{maxit}$.

The Implicit Euler's Method

Consider the initial value problem (IVP)

$$x(t_a) = x_a \quad (10a)$$

$$\dot{x}(t) = f(t, x(t)) \quad t_a \leq t \leq t_b \quad (10b)$$

It has the solution

$$x_{k+1} - x_k = \int_{x_k}^{x_{k+1}} dx = \int_{t_k}^{t_{k+1}} f(t, x(t)) dt \quad (11)$$

using $x_k = x(t_k)$. The integral $\int_{t_k}^{t_{k+1}} f(t, x(t)) dt$ can be approximated using the right-side-evaluation. This gives

$$x_{k+1} = x_k + (t_{k+1} - t_k) f(t_{k+1}, x_{k+1}) \quad (12)$$

Let $\Delta t = t_{k+1} - t_k$. Then

$$x_{k+1} = x_k + \Delta t f(t_{k+1}, x_{k+1}) \quad (13)$$

which can be rearranged to

$$R(x_{k+1}) = x_{k+1} - \Delta t f(t_{k+1}, x_{k+1}) - x_k = 0 \quad (14)$$

The Implicit Euler's Method

Consider the initial value problem (IVP)

$$x(t_a) = x_a \quad (15a)$$

$$\dot{x}(t) = f(t, x(t)) \quad t_a \leq t \leq t_b \quad (15b)$$

The finite-difference approximation

$$\frac{x_{k+1} - x_k}{\Delta t} \approx \frac{dx}{dt}(t_{k+1}) = \dot{x}(t_{k+1}) = f(t_{k+1}, x_{k+1}) \quad (16)$$

can be rearranged as

$$x_{k+1} = x_k + \Delta t f(t_{k+1}, x_{k+1}) \quad (17)$$

which is equivalent to

$$R(x_{k+1}) = x_{k+1} - \Delta t f(t_{k+1}, x_{k+1}) - x_k = 0 \quad (18)$$

Newton's method in the implicit Euler method

Nonlinear residual equation

$$R(x_{k+1}) = x_{k+1} - \Delta t f(t_{k+1}, x_{k+1}) - x_k = 0 \quad (19)$$

Jacobian

$$M = \frac{\partial R}{\partial x}(x_{k+1}) = I - \Delta t \frac{\partial f}{\partial x}(t_{k+1}, x_{k+1}) \quad (20)$$

Iterations in Newton's method

$$M \Delta x_{k+1} = R(x_{k+1}^{[l]}) \quad (21a)$$

$$x_{k+1}^{[l+1]} = x_{k+1}^{[l]} - \Delta x \quad (21b)$$

Stopping criterion: $\|R(x)\|_{\infty} \leq \epsilon$

Initial guess (explicit Euler):

$$x_{k+1}^{[0]} = x_k + \Delta t f(t_k, x_k) \quad (22)$$

Newton's Method in the Implicit Euler Method - Matlab

```
1 function x = NewtonsMethodODE(FunJac, tk, xk, dt, xinit, tol, maxit, varargin)
2
3 k = 0;
4 t = tk + dt;
5 x = xinit;
6 [f,J] = feval(FunJac,t,x,varargin{:})
7 R = x - f*dt - xk;
8 I = eye(length(xk));
9 while( (k < maxit) & (norm(R,'inf') > tol) )
10     k = k+1;
11     dRdx = I - J*dt;
12     dx = dRdx\R;
13     x = x - dx;
14     [f,J] = feval(FunJac,t,x,varargin{:});
15     R = x - dt*f - xk;
16 end
```

The Implicit Euler Method - A first crude implementation

```
1 function [T,X] = ImplicitEulerFixedStepSize(funJac,ta,tb,N,xa,varargin)
2
3 % Compute step size and allocate memory
4 dt = (tb-ta)/N;
5 nx = size(xa,1);
6 X = zeros(nx,N+1);
7 T = zeros(1,N+1);
8
9 tol = 1.0e-8;
10 maxit = 100;
11
12 % Eulers Implicit Method
13 T(:,1) = ta;
14 X(:,1) = xa;
15 for k=1:N
16     f = feval(fun,T(k),X(:,k),varargin{:});
17     T(:,k+1) = T(:,k) + dt;
18     xinit = X(:,k) + f*dt;
19     X(:,k+1) = NewtonsMethodODE(funJac,...
20                               T(:,k), X(:,k), dt, xinit, tol, maxit, varargin{:});
21 end
22
23 % Form a nice table for the result
24 T = T';
25 X = X';
```

Exercises

- ▶ Consider the van der Pol problem.
 1. Implement a function that returns the f and $J = \partial f / \partial x$:
`function [f,J] = VanderPolfunjac(t,x,mu)`
 2. Solve the problem using the explicit Euler method
 3. Solve the problem using the implicit Euler method (make sure you choose a time step that is small enough for the Newton method to converge)
 4. Compare the solutions