Queens College, CUNY, Department of Computer Science Numerical Methods CSCI 361 / 761 Fall 2017

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18 Lecture 18

Numerical solution of systems of ordinary differential equations

- In this lecture we continue the study of **initial value problems**.
- In this lecture we present a **semi-implicit** integration algorithm.

18.1 Basic notation

- We repeat the basic definitions from the previous lecture.
- Let the system of coupled differential equations be

$$\frac{d\mathbf{y}}{dx} = \mathbf{f}(x, \mathbf{y}). \tag{18.1.1}$$

- There are n unknown variables y_j , j = 1, ..., n.
- The starting point is $x = x_0$, and the initial value y_0 is given.
- The above is called the Cauchy problem or initial value problem.
- Our interest is to integrate eq. (18.1.1) numerically, using steps h_i , so $x_{i+1} = x_i + h_i$.
- The steps h_i need not be of equal size.
- We define $y_i = y(x_i)$.
- We employ the notation $\boldsymbol{y}_i^{\mathrm{ex}} = \boldsymbol{y}^{\mathrm{ex}}(x_i)$ to denote the exact solution.
- We employ the notation $\boldsymbol{y}_i^{\text{num}} = \boldsymbol{y}^{\text{num}}(x_i)$ to denote the numerical solution.

18.2 Semi-implicit method

- The system of equations is given by eq. (18.1.1).
- At the step i we employ a finite difference to approximate the derivative, which yields the implicit equation

$$\mathbf{y}_{i+1}^{\text{imp}} = \mathbf{y}_{i}^{\text{imp}} + h_{i} \mathbf{f}(x_{i+1}, \mathbf{y}_{i+1}^{\text{imp}}).$$
 (18.2.1)

- We employ a multivariate Newton-Raphson algorithm to solve eq. (18.2.1).
- For the first iteration, we write

$$f(x_{i+1}, \mathbf{y}_{i+1}^{\text{imp}}) = f(x_i, \mathbf{y}_i^{\text{imp}}) + \frac{\partial f}{\partial \mathbf{y}} \cdot (\mathbf{y}_{i+1}^{\text{imp}} - \mathbf{y}_i^{\text{imp}}).$$
(18.2.2)

• The matrix $\partial f/\partial y$ is the **Jacobian matrix** J

$$J_{jk} = \frac{\partial f_j}{\partial y_k} \qquad (1 \le j, k \le n). \tag{18.2.3}$$

• We substitute into eq. (18.2.1) to obtain

$$\mathbf{y}_{i+1}^{\text{imp}} - \mathbf{y}_{i}^{\text{imp}} = h_{i} \left[\mathbf{f}(x_{i}, \mathbf{y}_{i}^{\text{imp}}) + J(\mathbf{y}_{i+1}^{\text{imp}} - \mathbf{y}_{i}^{\text{imp}}) \right].$$
 (18.2.4)

• Rearranging the terms in eq. (18.2.4) yields a matrix equation in n variables (I is the unit $n \times n$ matrix)

$$(I-J)\boldsymbol{y}_{i+1}^{\text{imp}} = (I-J)\boldsymbol{y}_{i}^{\text{imp}} + h_{i}\boldsymbol{f}(x_{i},\boldsymbol{y}_{i}^{\text{imp}}).$$

$$(18.2.5)$$

- The formula in eq. (18.2.5) is called a **semi-implicit method**.
- We stop the Newton–Raphson iteration after only one iteration.
- To compute y_{i+1}^{imp} , eq. (18.2.5) must be solved using a matrix equation solver.
 - 1. We can write the formal solution of eq. (18.2.5) as

$$\mathbf{y}_{i+1}^{\text{imp}} = \mathbf{y}_{i}^{\text{imp}} + h_{i} (I - J)^{-1} \mathbf{f}(x_{i}, \mathbf{y}_{i}^{\text{imp}}).$$
 (18.2.6)

- 2. Note that eq. (18.2.5) is simply a one-iteration approximation to solve eq. (18.2.1).
- 3. The semi-implicit algorithm is not guaranteed to be computationally stable.
- 4. The basic hope in the semi-implicit algorithm is that if the magnitude of the stepsize $|h_i|$ is small enough, using only one Newton-Rahpson iteration will be satisfactory.