

Numerical Solution of ODEs

Basic Concepts

1. **DE** = Differential Equation: an equation for an unknown function involving the unknown and its derivatives

eg. $y' + y - e^t = 1 : y = y(t)$

$3xy'' + y^2 = 0 : y = y(x)$

$(t^2 + 1)y'' - 3yy' + 6t = 0 : y = y(t)$

$$\frac{\partial u}{\partial t} = D \frac{\partial^2 u}{\partial x^2} ; u = u(x, t)$$

$$\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = f(x, y) ; u = u(x, y)$$

Order of a DE is the order of highest derivative appearing in the DE.

ODE : DE for unknown of one independent variable: $y(t)$ M231, 431, 531

PDE : DE for unknown of many indep, variables: $u(x, y, t)$ M435, 475, 535-6, 635

Solution of a DE is a function that satisfies the DE at each t

eg. $y' = y + e^t$ has a solution $y(t) = Ce^t + te^t$ for any const. C

indeed: $y' = [Ce^t + te^t]' = Ce^t + e^t + te^t \equiv y + e^t \quad \forall C$

Some simple types of ODEs can be solved analytically (M231, 431-2)
all others only numerically!

Solution of n -th order ODE involves n arbitrary constants!

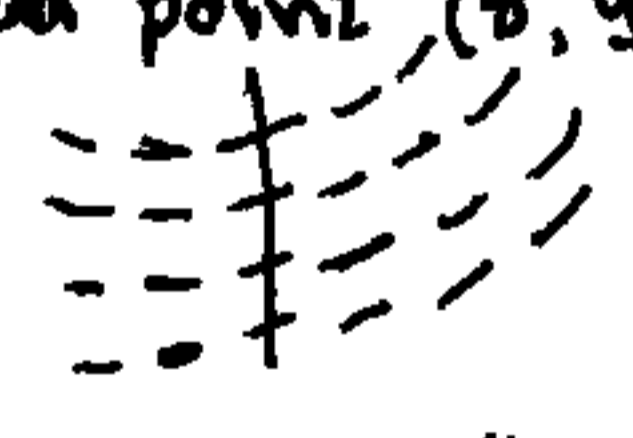
In general, a given DE has many solutions (or none)

so additional conditions must be specified in order to determine a single solution

e.g. $y' = y + e^t$ only specifies the slope $y'(t)$ at each point (t, y) .

a direction field on (t, y) plane:

but ∞ -ly many curves have such slopes



To pick one trajectory, must specify a "starting" point (t_0, y_0)
i.e. $y(t_0) = y_0$

2. We will consider only 1st order ODEs in standard form: $y' = f(t, y)$

with an initial condition: $y(t_0) = y_0$. Constitutes an Initial Value Problem

standard IVP for 1st order ODE: (IVP) $\begin{cases} y' = f(t, y), & t_0 \leq t \leq t_{end} \\ y(t_0) = y_0 \end{cases}$ ODE
IC

$\begin{cases} y' = f(t, y) \\ y(t_0) = y_0 \end{cases}$ is equivalent to the Integral Equ: $y(t) = y_0 + \int_{t_0}^t f(s, y(s)) ds$

3. Well-posedness

A problem is well-posed if

1. solution(s) exist (Existence property)
2. solution is unique (Uniqueness)
3. solution depends continuously on data (Continuous dep. on data)
(small changes in data \Rightarrow small change in solution)

else it is called ill-posed.

Inverse problems are ill-posed! (given a solution find the problem)

most important: parameter identification problems are ill-posed!

(from measurements of solution find physical properties)

Well-posedness Theorem for (IVP) $\begin{cases} y' = f(t, y) \\ y(t_0) = y_0 \end{cases}$

If $f(t, y)$ and $\frac{\partial f}{\partial y}(t, y)$ are continuous and bounded on a rectangle $R = [a, b] \times [c, d]$ of (t, y) plane; then

for any $(t_0, y_0) \in R$ the IVP is well-posed near the initial pt (t_0, y_0)

i.e. there exists unique solution $y(t)$ in some subinterval $(t_0 - r, t_0 + r), r >$ which depends continuously on the data.

If $\frac{\partial f}{\partial y} \gg 0$ then trajectories get further apart as $t \uparrow$, errors get amplified, ill-conditioned (IVP), equilibria are unstable

If $\frac{\partial f}{\partial y} < 0$ then trajectories get closer, errors are damped as $t \uparrow$, well-conditioned (IVP), equilibria are stable

Hard to tell which case because sign of $\frac{\partial f}{\partial y}$ depends on y , may be changing...

Typically, to test num. methods we apply them to prototype linear IVP: $\begin{cases} y' = \lambda y \\ y(0) = y_0 \end{cases}$
whose solution is $y(t) = y_0 e^{\lambda t}$

4. Systems of 1st order: (IVP)
$$\begin{cases} y_1' = f_1(t, y_1, y_2, \dots, y_n) \\ y_2' = f_2(t, y_1, \dots, y_n) \\ \vdots \\ y_n' = f_n(t, y_1, \dots, y_n) \end{cases}, \begin{cases} y_1(t_0) = y_1^0 \\ y_2(t_0) = y_2^0 \\ \vdots \\ y_n(t_0) = y_n^0 \end{cases}$$

can write it in vector form:

(IVP)
$$\begin{cases} \vec{y}' = \vec{F}(t, \vec{y}) \\ \vec{y}(t_0) = \vec{y}^0 \end{cases} \text{ where } \vec{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}, \vec{F} = \begin{bmatrix} f_1 \\ \vdots \\ f_n \end{bmatrix}, \vec{y}^0 = \begin{bmatrix} y_1^0 \\ \vdots \\ y_n^0 \end{bmatrix}$$

All methods developed for scalar case (single ODE) also apply to vector case and all ODE solver software are for systems.

5. Higher order IVPs can be written as 1st order systems:

(IVP)
$$\begin{cases} y^{(n)} = g(t, y, y', \dots, y^{(n-1)}) \\ y(t_0) = y_0, y'(t_0) = y_1^0, \dots, y^{(n-1)}(t_0) = y_{n-1}^0 \end{cases}$$

set $w_1 = y, w_2 = y', \dots, w_n = y^{(n-1)}$, then

(IVP)
$$\begin{cases} w_1' = w_2, & w_1(t_0) = y_0 \\ w_2' = w_3, & w_2(t_0) = y_1^0 \\ \vdots \\ w_{n-1}' = w_n, & w_{n-1}(t_0) = y_{n-2}^0 \\ w_n' = g(t, w_1, \dots, w_n), & w_n(t_0) = y_{n-1}^0 \end{cases}$$

i.e.
$$\begin{cases} \vec{w}' = \vec{F}(t, \vec{w}) \\ \vec{w}(t_0) = \vec{w}^0 = \vec{y}^0 \end{cases}$$

e.g. $3xy'' + y^2 = 0$: set $w_1 = y, w_2 = y' \Rightarrow \begin{cases} w_1' = w_2 \\ w_2' = y'' = -\frac{y^2}{3x} = -\frac{w_1^2}{3x} \end{cases}$

or
$$\vec{w}' = \begin{bmatrix} w_1' \\ w_2' \end{bmatrix} = \begin{bmatrix} w_2 \\ -\frac{w_1^2}{3x} \end{bmatrix} =: \vec{F}(t, \vec{w}) \quad 2 \times 2 \text{ system}$$

4

Numerical solutions for (IVP) $\begin{cases} y' = f(t, y), t_0 < t < t_{\text{end}} \\ y(t_0) = y_0 \end{cases}$

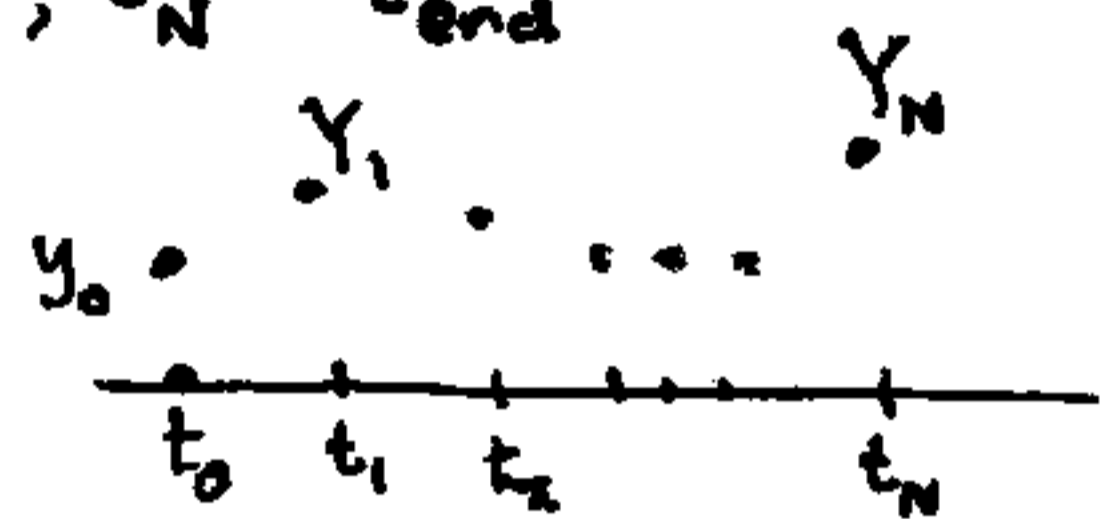
By the Well-posedness Thm, if $f, \frac{\partial f}{\partial y}$ are cont's and bounded on some \mathcal{R} , then the IVP is well-posed (at least locally).

Exact analytic solution are possible only for very particular $f(t, y)$, studied in M231. All others must be approximated numerically.

Numerical solution by time-discretization:

Discretize time interval $[t_0, t_{\text{end}}]$ into N subintervals by (usually) equispaced nodes $t_0 < t_1, \dots, t_N = t_{\text{end}}$ with time-steps $\Delta t = \frac{t_{\text{end}} - t_0}{N} = h$

$$t_n = t_0 + n \cdot \Delta t, \quad n = 0:N$$



and seek approximations

$$Y_n \approx y(t_n), \quad n = 0:N \quad \text{only at the } N \text{ times } t_n$$

with $Y_0 = y_0 = y(t_0)$ from IC.

Numerical schemes produce values $Y_0, Y_1, \dots, Y_n, \dots, Y_N$ by time-stepping.

discretization error $= y(t_n) - Y_n$ should $\rightarrow 0$ as $\Delta t \rightarrow 0$ (more later...)

There are many schemes, and more are being developed still...

Dictum: There is no best method for all problems!

Numerical methods should be:

consistent, convergent, stable (more later)
and efficient and robust

Computational cost: number of $f(t_n, Y_n)$ evaluations needed to get to t_{end}