Chapter 5: Numerical Solutions of Ordinary Differential Equations

Prof. Alok Shukla

Department of Physics IIT Bombay, Powai, Mumbai 400076

Course Name: Introduction to Numerical Analysis (PH 307)

Ordinary Differential Equations

- In this chapter we will explore various numerical methods aimed at numerically solving ordinary differential equations (ODEs)
- But, what is an ODE, and how does it differ from a partial differential equation (PDE)?
- A PDE is a differential equation involving more than one variable
- On the other hand, an ODE is a differential equation involving just one variable, taken normally to be x
- An ODE can be considered as a special case of a PDE in which the number of variables is just one.
- Therefore, we will study the numerical solutions of ODEs before tackling the PDEs



ODEs (contd.)

- For the time being we will focus on linear differential equations
- A general n-th order linear differential equation for an unknown function y(x) can be written as

$$\left\{D^n + G_{n-1}(x)D^{n-1} + \dots + G_1(x)D + G_0(x)\right\}y = Q(x), \quad (1)$$

where $D^n = \frac{d^n}{dx^n}$ and $G_i(x)$, i = 0, ..., n-1, Q(x) are known functions of the variable x.

• Given the ODE of Eq. 1, and the values of y(x) and/or its derivatives at some values of x, one has to solve for the unknown function y(x) for all values of x for which the ODE is defined.

First order initial value problems

- As far as the numerical methods are concerned, we will show that the methods developed for the first-order ODE (n=1) can be used to solve a large class of higher-order ODEs
- ullet Note that for n=1, Eq. 1, reduces to an intial-value problem

$$\frac{dy}{dx} = F(x, y)$$

$$y(x_0) = y_0 \text{ (given)}$$
(2)

above F(x,y) is a function of both x and y, and the value of the solution y(x) is given to be y_0 at a point $x = x_0$

• y_0 is also called the initial value, hence the name "initial value problem" for such ODEs.

Numerical solution

- Let us assume that we want to numerically determine y(x), for all values of $x \in [a,b]$
- As in earlier cases of interpolation and numerical integration, we divide this interval in N equal bins whose width h is given by

$$h = \frac{b - a}{N}$$

So that

$$x_0 = a$$

$$x_N = b$$

$$x_{n+1} = x_n + h$$

• With the initial condition $y_0 = y(x_0) = y(a)$



Numerical solution: Euler's method

- Euler's method is one of the simplest techniques for solving a first-order ODE numerically
- Because of its simplicity, it is not very accurate.
- Nevertheless, it is worthwhile to start the discussion of the numerical methods from this approach
- As it turns out, it belongs to a class of approaches called Taylor Expansion Approach
- Let us assume that we have determined the solution for the x values $\{x_0, x_1, x_2, \dots, x_n\}$
- This implies that we know the values $\{y_0, y_1, y_2, \dots, y_n\}$, where $y_i = y(x_i)$



Euler's Method (contd.)

- Using this we want to compute the value of the function y_{n+1} at the next point x_{n+1}
- We can write

$$y_{n+1} = y_n + \int_{x_n}^{x_{n+1}} y'(x) dx,$$
 (3)

where $y'(x) = \frac{dy}{dx}$

For small values of h, we can approximate

$$\int_{x_n}^{x_{n+1}} y'(x) dx \approx h y_n'$$

Because our ODE is

$$y'(x) = F(x,y)$$

we can substitute $y'_n = F(x_n, y_n)$ above leading to the famous Euler's formula

$$y_{n+1} = y_n + hF(x_n, y_n)$$
 (4)

Euler's method...

- Note that this formula has the form of a recursion relation, and it yields the solution at the point $x = x_{n+1}$ in terms of solution at the previous point $x = x_n$
- Because the solution at $x = x_0$ is known, therefore, this formula will yield the solution for all values of x_i , i = 1, ..., N

Euler's Method: An Example

 Let us consider a simple ODE whose exact solution is known, namely

$$\frac{dy}{dx} = y$$

$$y_0 = y(x = 0) = 1$$
(5)

• We know that the exact solution of this initial value problem for $x \in [0, \infty)$ is

$$y(x) = e^x$$

- Let us try to obtain numerical solution of this ODE using the Euler's method in the domain $x \in (0,1)$
- The bin size h will be

$$h = \frac{b-a}{N} = \frac{1}{N}$$

$$x_0 = 0, x_N = 1$$

$$x_n = nh = \frac{n}{N} \text{ for } 0 \le n \le N$$

Euler's Method: Example (contd.)

• Clearly, here F(x,y) = y, therefore on substituting in Eq. 4, we obtain

$$y_{n+1} = y_n + hy_n = (h+1)y_n \tag{6}$$

ullet For the present case of the range of solution $x\in[0,1]$, we have

$$y_{n+1} = \left(\frac{N+1}{N}\right) y_n \tag{7}$$

Euler Method: Example...

• If we take N = 10, we have

$$y_{n+1} = \left(\frac{11}{10}\right) y_n = 1.1 y_n$$

• This formula readily yields

Xn	$y_n = (1.1)^n$ (Euler)	$y_n = e^{x_n}$ (exact)
0.0	1.000	1.0
0.1	1.100	1.105
0.2	1.210	1.221
:	;	:
0.9	2.358	2.450
1.0	2.594	2.718

Euler Method: Example

- We note that with the increasing values of x, the Euler solution is systematically smaller than the exact one
- Q: Why is that the case?
 - The first reason is that we have used a rather large value of h=0.1. If we were to use N=100, so that h=0.01. Using that in Eq. 7, obtain $y_n=(1.01)^n$ leading to

$$y(x = 0.1) = (1.01)^{10} = 1.105$$

 $y(x = 0.9) = (1.01)^{90} = 2.445$
 $y(x = 1.0) = (1.01)^{100} = 2.705$ (8)

- Now we note that there is much better agreement between the exact values and the calculated values
- But, if we do calculations for larger values of x, even for this value of h, the disagreement will become severe.
- Therefore, a smaller value of h will lead to a better agreement with the exact values because as we will see that the errors in the Euler's method scale as $\sim h^2$



Problems with the Euler's method

- But, there are some inherent problems in the Euler's approach.
- We will show that it is the lowest order formula obtained from the general Taylor expansion approach for solving the ODEs
- Another problem is that we are using the slope of y(x) at the point $x = x_n$ to compute it at $x = x_{n+1}$
- What if we used its average value over x_n and x_{n+1}

$$y_{n+1} = y_n + \frac{h}{2} \left(y'_n + y'_{n+1} \right) \tag{9}$$

- The problem with this formula is that it is trying to determine y_{n+1} in terms of y'_{n+1} , which we don't know!
- However, such formulas are used in what are called the Predictor-Corrector methods which we will discuss later
- Next, we discuss the Taylor expansion approach for solving the first-order ODEs

ODEs



The Taylor Expansion Approach

• Let us Taylor expand $y_{n+1} = y(x_n + h)$ around the point x_n

$$y_{n+1} = y_n + hy_n' + \frac{h^2}{2}y_n'' + \frac{h^3}{6}y_n''' + \frac{h^4}{24}y_n'''' + \cdots$$
 (10)

- Note that if we discard all the terms except the first order one in h, we obtain the Euler's formula of Eq. 4.
- We note that this approach is systematically improvable by retaining the successively higher-order terms
- However, the problem is how to calculate the second and higher-order derivatives needed in Eq. 10, given y'(x) = F(x,y)
- Mathematically, speaking it can be done as follows

$$y''(x) = \frac{dy'}{dx} = \frac{dF(x,y)}{dx}$$
$$= \frac{\partial F}{\partial x} + \frac{\partial F}{\partial y} \frac{dy}{dx} = F_x + FF_y$$

above we have used the notations $F_x = \frac{\partial F}{\partial x}$ and $F_y = \frac{\partial F}{\partial y_0}$

Taylor Expansion approach...

Similarly

$$y'''(x) = \frac{dy''}{dx} = \frac{d}{dx} (F_x + FF_y)$$

$$= F_{xx} + F_{xy}F + (F_x + FF_y)F_y + FF_{yx} + FF_{yy}F$$

$$= F_{xx} + 2FF_{xy} + F^2F_{yy} + F_xF_y + FF_y^2$$

above we used the fact that $F_{xy} = F_{yx}$

- We note that the calculation of the higher order derivatives becomes quite tedious
- However, if it can be done for some ODE, this approach yields quite accurate results

Taylor Expansion Method; Example revisited

• Let us again consider the ODE of Eq. 5

$$\frac{dy}{dx} = y$$
$$y_0 = y(x = 0) = 1$$

• The Taylor expansion solution for this will be

$$y_{n+1} = (1 + h + \frac{h^2}{2} + \frac{h^3}{6} + \frac{h^4}{24} + \cdots)y_n$$

• Let us again consider h=0.1, which leads to

$$y_{n+1} = (1+0.1+0.005+0.00017+0.0000042+\cdots)y_n$$

 $\implies y_n = (1.105)^n \text{ (2nd order)}$
 $y_n = (1.1051666...)^n \text{ (3rd order)}$
 $y_n = (1.1051708332..)^n \text{ (4th order)}$

• We tabulate the results obtained for various orders of Taylor expansion for some points $x \in [0,1]$, and compare those to the exact results

Taylor Expansion Approach: example...

,						
	x_n	Уn	Уn	Уn	Уn	$y_n = e^{x_n}$
		(Euler)	(2nd order)	(3rd order)	4th order)	(exact)
	0.0	1.0000	1.0000	1.0000	1.0000	1.0000
	0.1	1.1000	1.1050	1.1052	1.1052	1.1052
	0.2	1.2100	1.2210	1.2214	1.2214	1.2214
	!	:		:	:	÷
	0.9	2.3579	2.4562	2.4595	2.4596	2.4596
	1.0	2.5937	2.7141	2.7182	2.7183	2.7183

• Note that the 4th-order result is exact up to 10^{-4} , which it should be for $h = 10^{-1}$.

Runge-Kutta Approach

 We say that the Taylor-expansion approach for solving the first order ODEs of the form

$$y' = F(x,y)$$
, with $y(x_0) = y_0$

is based on the formula

$$y_{n+1} = y_n + hy'_n + \frac{h^2}{2}y''_n + \frac{h^3}{6}y'''_n + \frac{h^4}{24}y'''_n + \cdots$$

- We also briefly discussed that it can be tedious to calculate the higher order derivatives $y^{(r)}(x)$ $(r \ge 2)$ needed for the Taylor expansion
- Runge, Kutta, and a few others proposed an approach to avoid this problem approach is aimed at avoiding this by using a formula of the form

$$y_{n+1} = y_n + \alpha_0 k_0 + \alpha_1 k_1 + \dots + \alpha_p k_p$$
 (11)



Runge-Kutta Approach...

where

$$k_{0} = hF(x_{n}, y_{n})$$

$$k_{1} = hF(x_{n} + \mu_{1}h, y_{n} + \lambda_{10}k_{0})$$

$$k_{2} = hF(x_{n} + \mu_{2}h, y_{n} + \lambda_{20}k_{0} + \lambda_{21}k_{1})$$

$$\vdots = \vdots$$

$$k_{p} = hF(x_{n} + \mu_{p}h, y_{n} + \lambda_{p0}k_{0} + \lambda_{p1}k_{1} + \dots + \lambda_{pp-1}k_{p-1})$$
(12)

the unknown coefficients α_i , μ_i , and λ_{ij} are determined by Taylor expanding the above equations around x_n and y_n , and then comparing the results with Taylor expansion of Eq. 10 to a suitable order.

• Next, we do that for the simple case of p=1

Verification of the second-order Runge-Kutta Formula

ullet For p=1, the Runge-Kutta (RK) equations are

$$y_{n+1} = y_n + \alpha_0 k_0 + \alpha_1 k_1$$

$$k_0 = hF(x_n, y_n)$$

$$k_1 = hF(x_n + \mu h, y_n + \lambda k_0)$$
(13)

• Next, we Taylor expand k_1 of the previous equation around points (x_n, y_n) retaining terms up to $O(h^2)$, using the notations $F = F(x_n, y_n)$, $F_x = \frac{\partial F}{\partial x}$, and $F_y = \frac{\partial F}{\partial y}$,

$$k_1 = h \left[F + h \mu F_x + \lambda k_0 F_y + O(h^2) \right] = h F + h^2 \mu F_x + \lambda h^2 F F_y + O(h^3)$$

• Substituting the expressions for k_0 and k_1 in the first of Eq. 13

$$y_{n+1} = y_n + (\alpha_0 + \alpha_1)hF + h^2\alpha_1(\mu F_x + \lambda FF_y) + O(h^3)$$
 (14)



RK2 derivation (contd.)

• If we apply, the 2nd-order Taylor expansion approach, we have

$$y_{n+1} = y_n + hy'_n + \frac{h^2}{2}y''_n + O(h^3).$$

Using the fact

$$y_{n}^{'} = F = F(x_{n}, y_{n})$$
$$y_{n}^{''} = F_{x} + FF_{y}$$

substituting these above, we obtain

$$y_{n+1} = y_n + hF + \frac{h^2}{2} (F_x + FF_y) + O(h^3)$$
 (15)

RK2-contd...

• Comparing different powers of h in Eqs. 14 and 15, we obtain

$$\alpha_0 + \alpha_1 = 1$$

$$\mu \alpha_1 = \frac{1}{2}$$

$$\lambda \alpha_1 = \frac{1}{2}$$

• These equations can be satisfied for any constant c

$$\alpha_0 = 1 - c$$

$$\alpha_1 = c$$

$$\mu = \frac{1}{2c}$$

$$\lambda = \frac{1}{2c}$$

RK2 Final Formula

• A popular choice is $c=\frac{1}{2}$, leading to the famous symmetric RK2 formula

$$y_{n+1} = y_n + \frac{1}{2}(k_0 + k_1)$$

$$k_0 = hF(x_n, y_n)$$

$$k_1 = hF(x_n + h, y_n + k_0)$$
(16)

Higher Order RK Formulas: RK-3

- One can similarly derive higher-order RK formulas
- Several RK-3 (third order) formulas are possible, while one popular choice is

$$y_{n+1} = y_n + \frac{1}{6} (k_0 + 4k_1 + k_2) + O(h^4)$$

$$k_0 = hF(x_n, y_n)$$

$$k_1 = hF\left(x_n + \frac{h}{2}, y_n + \frac{k_0}{2}\right)$$

$$k_2 = hF(x_n + h, y_n + 2k_1 - k_0)$$
(17)

Fourth-Order RK Formula: RK-4

• A fourth-order Runge-Kutta formula is given by

$$y_{n+1} = y_n + \frac{1}{6} (k_0 + 2k_1 + 2k_2 + k_3) + O(h^5)$$

$$k_0 = hF(x_n, y_n)$$

$$k_1 = hF\left(x_n + \frac{h}{2}, y_n + \frac{k_0}{2}\right)$$

$$k_2 = hF\left(x_n + \frac{h}{2}, y_n + \frac{k_1}{2}\right)$$

$$k_3 = hF(x_n + h, y_n + k_2)$$
(18)

Predictor-Corrector Methods

- In this course we will not be going deep into the topic of Predictor-Corrector (PC) methods,
- This is because more accurate results can be obtained using simpler algorithms by adopting methods like Taylor series expansion and RK-n
- However, we will briefly mention the lowest order PC method namely Euler PC approach
- In the PC approach, one predicts the value of the solution at a point x_n using the predictor formula
- Next, it is corrected using the so-called corrector formula
- This cycle is repeated for each point x_n until the results of predictor and corrector agree with each other



Euler Predictor-Corrector Method

For the Euler PC approach, the relevant formulas are

$$y_{n+1} = y_n + hy'_n$$
 Predictor Formula
 $y_{n+1} = y_n + \frac{h}{2} \left(y'_{n+1} + y'_n \right)$ Corrector Formula (19)

- Note that the value of $y_{n+1}^{'}$ needed in the corrector formula is obtained using the ODE $y_{n+1}^{'} = F(x_{n+1}, y_{n+1})$, where the value of y_{n+1} obtained from the predictor formula is used
- This cycle is repeated for each x_n until the values of y_{n+1} obtained from the predictor and the corrector formulas agree with each other

Systems of Ordinary Differential Equations

Let us consider a system of p first-order ODEs

$$y_{i}^{'} = F_{i}(x, y_{1}, y_{2}, ..., y_{p})$$

 $y_{i}(x_{0}) = y_{i0}$ $i = 1, 2, 3, ..., p$

Let us adopt the vector notations

$$Y = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_p \end{pmatrix}, \quad \tilde{F} = \begin{pmatrix} F_1 \\ F_2 \\ F_3 \\ \vdots \\ F_p \end{pmatrix}$$

System of ODEs: An Example

 Using the vector notation, the system of ODEs can be written in the compact form

$$Y' = \tilde{F},$$

subect to the initial conditions

$$Y_0 = Y(x_0) = \begin{pmatrix} y_1(x_0) \\ y_2(x_0) \\ y_3(x_0) \\ \vdots \\ y_p(x_0) \end{pmatrix}$$

• Let us consider a simple example with p=2

$$y_{1}^{'} = -y_{1} - y_{2} = F_{1}(x, y_{1}, y_{2})$$

$$y_{2}^{'} = y_{1} - y_{2} = F_{2}(x, y_{1}, y_{2})$$
(20)

with the initial conditions $y_1(0) = 1$, $y_2(0) = 0$



A system of ODEs: Taylor Expansion Approach

Clearly, this equations can be put in the matrix form

$$Y' = AY$$

where

$$Y = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$$
 and $A = \begin{pmatrix} -1 & -1 \\ 1 & -1 \end{pmatrix}$

- Let us set up the Taylor Expansion approach for this system of ODEs
- We will have

$$y_1^{n+1} = y_1(x_n + h) = y_1^n + hy_1^{'n} + \frac{h^2}{2}y_1^{''n} + \frac{h^3}{6}y_1^{'''n} + \frac{h^4}{24}y_1^{''''n} + \cdots$$

$$y_2^{n+1} = y_2(x_n + h) = y_2^n + hy_2^{'n} + \frac{h^2}{2}y_2^{''n} + \frac{h^3}{6}y_2^{'''n} + \frac{h^4}{24}y_2^{''''n} + \cdots$$
(21)

ODEs



A system of ODEs: Taylor...

Using, Eqs. 20, Let us calculate the higher order derivatives

$$y_{1}^{"} = \frac{dy_{1}^{'}}{dx} = \frac{dF_{1}(x, y_{1}, y_{2})}{dx} = F_{1x} + F_{1y_{1}}y_{1}^{'} + F_{1y_{2}}y_{2}^{'}$$

$$= F_{1x} + F_{1y_{1}}F_{1} + F_{1y_{2}}F_{2} = 0 - (-y_{1} - y_{2}) - (y_{1} - y_{2})$$

$$= 2y_{2}$$

$$y_{1}^{"'} = \frac{dy_{1}^{"}}{dx} = 2y_{2}^{'} = 2y_{1} - 2y_{2}$$

Similarly

$$y_{2}^{"} = \frac{dy_{2}^{'}}{dx} = \frac{dF_{2}(x, y_{1}, y_{2})}{dx} = F_{2x} + F_{2y_{1}}y_{1}^{'} + F_{2y_{2}}y_{2}^{'}$$

$$= F_{2x} + F_{2y_{1}}F_{1} + F_{2y_{2}}F_{2} = 0 + (-y_{1} - y_{2}) - (y_{1} - y_{2})$$

$$= -2y_{1}$$

$$y_{2}^{"'} = \frac{dy_{2}^{"}}{dx} = -2y_{1}^{'} = -2(-y_{1} - y_{2}) = 2y_{1} + 2y_{2}$$

System of ODEs: Taylor Expansion...

• Therefore, using Eqs 21, and the expressions for derivatives of y_1/y_2 , the 3rd-order Taylor expansion solution for this system can be written as

$$y_1^{n+1} = y_1^n - hy_1^n - hy_2^n + h^2y_2^n + \frac{h^3}{3}y_1^n - \frac{h^3}{3}y_2^n + \cdots$$
$$y_2^{n+1} = y_2^n + hy_1^n - hy_2^n - h^2y_1^n + \frac{h^3}{3}y_1^n + \frac{h^3}{3}y_2^n + \cdots$$

- The values of y_1^{n+1}/y_2^{n+1} can be easily generated from these expansions, starting with n=0.
- For this particular case, extension to higher orders of expansion is quite trivial

System of ODEs: Runge-Kutta Approach

- Extension of the Runge-Kutta approach to a system of equations is almost trivial
- It basically amounts to applying the RK method to each ODE of the system, individually
- For the general case of p coupled ODEs, the RK-4 formulas will be, for i = 1, 2, 3, ..., p

$$y_{i}^{n+1} = y_{i}^{n} + \frac{1}{6} \left(k_{0}^{i} + 2k_{1}^{i} + 2k_{2}^{i} + k_{3}^{i} \right)$$

$$k_{0}^{i} = hF_{i}(x_{n}, y_{1}^{n}, y_{2}^{n}, \dots, y_{p}^{n})$$

$$k_{1}^{i} = hF_{i} \left(x_{n} + \frac{h}{2}, y_{1}^{n} + \frac{k_{0}^{1}}{2}, y_{2}^{n} + \frac{k_{0}^{2}}{2}, \dots, y_{p}^{n} + \frac{k_{0}^{p}}{2} \right)$$

$$k_{2}^{i} = hF_{i} \left(x_{n} + \frac{h}{2}, y_{1}^{n} + \frac{k_{1}^{1}}{2}, y_{2}^{n} + \frac{k_{1}^{2}}{2}, \dots, y_{p}^{n} + \frac{k_{1}^{p}}{2} \right)$$

$$k_{3}^{i} = hF_{i} \left(x_{n} + h, y_{1}^{n} + k_{2}^{1}, y_{2}^{n} + k_{2}^{2}, \dots, y_{p}^{n} + k_{2}^{p} \right)$$

Higher-Order Ordinary Differential Equations

- We can very easily transform and p-th order initial value problem ODE into a set of p first order coupled first-order ODEs
- Then the pth-order ODE can be solved using the numerical methods developed in the previous section for solving the system of first-order ODEs
- At the beginning of this chapter, we considered a general nth-order ODE

$$\left\{y^{(p)} + G_{n-1}(x)y^{(p-1)} + \dots + G_1(x)y^{(1)} + G_0(x)y\right\} = Q(x),$$
(23)

where $y^{(i)} = \frac{d^i y}{dx^i}$, etc., where the initial conditions $y(x_0) = y_0$, $y^{(1)}(x_0) = y_0^1$, $y^{(2)}(x_0) = y_0^2$,..., $y^{(p-1)}(x_0) = y_0^p$ are specified.

We can rewrite Eq. 23 as

$$y^{(p)}(x) = F(x, y, y^{(1)}, y^{(2)}, \dots, y^{(p-1)}),$$
 (24)

where the function F will depend on the exact form of the Eq.

23.



Higher order ODEs...

• Next, we define the following transformations

$$y_{1} = y(x)$$

$$y_{2} = y^{(1)}(x)$$

$$y_{3} = y^{(2)}(x)$$

$$\vdots = \vdots$$

$$y_{p-1} = y^{(p-2)}(x)$$

$$y_{p} = y^{(p-1)}(x)$$
(25)

• Using these definitions, we can write Eq. 24 as

$$y^{p'}(x) = F(x, y_1, y_2, ..., y_p),$$
 (26)

Higher order ODEs...

 We can combine Eqs. 25 and 26 to define the set of n coupled first-order ODEs

$$y'_{1}(x) = f_{1}(x, y_{1}, y_{2}, ..., y_{p})$$

$$y^{2'}(x) = f_{2}(x, y_{1}, y_{2}, ..., y_{p})$$

$$\vdots = \vdots$$

$$y'_{p-1}(x) = f_{p-1}(x, y_{1}, y_{2}, ..., y_{p})$$

$$y'_{p}(x) = f_{p}(x, y_{1}, y_{2}, ..., y_{p})$$
(27)

where

$$f_i(x, y_1, y_2, \dots, y_p) = y_{i+1},$$
 for $i = 1, 2, 3, \dots, n-1$
 $f_p(x, y_1, y_2, \dots, y_p) = F(x, y_1, y_2, \dots, y_p)$ (28)

Second Order ODE: Examples

 Let us first consider the differential equation of a simple-harmonic oscillator of unit angular frequency

$$\frac{d^2x}{dt^2} + x = 0$$

$$x(0) = 0$$

$$x'(0) = 1,$$
(29)

we know the exact solution of this initial value problem is $x(t) = \sin t$. If we instead change the initial conditions to x(0) = 1 and x'(0) = 0, the exact solution becomes $x(t) = \cos t$.

 We can easily convert Eq. 29 into two coupled linear ODEs using the transformations

$$x_1 = x$$
$$x_2 = \frac{dx}{dt}$$

Second-order ODEs: examples

With this, the pair of coupled equations becomes

$$x_{1}^{'} = f_{1}(t, x_{1}, x_{2}) = x_{2}$$

 $x_{2}^{'} = f_{2}(t, x_{1}, x_{2}) = -x_{1}$

with the initial conditions $x_1(0) = 0$, $x_2(0) = 1$.

- This equation system can be solved quite easily using the methods described in the previous section
- Next, consider the second-order differential equation

$$y'' + y' + y^2 = x$$
, $y(0) = 1$, $y'(0) = 0$

We make the transformation

$$y_1 = y$$
$$y_2 = y'$$

Second-Order ODEs: Examples...

• With this, our coupled equations become

$$y_1' = f_1(x, y_1, y_2) = y_2$$

 $y_2' = f_2(x, y_1, y_2) = x - y_1^2 - y_2$

which can be solved using the initial conditions $y_1(0) = 1$, $y_2(0) = 0$