

# COMPUTATIONAL PHYSICS

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# Ordinary differential equations

- Initial-value problems
- The Euler methods
- Predictor-corrector methods
- The Runge-Kutta method
- Chaotic dynamics
- Boundary-value problems
- The shooting method
- Linear equations
- Eigenvalue problems

Most problems in physics and engineering appear in the form of **differential equations**.

For example

① the motion of a **classical** particle is described by **Newton's equation**

$$\vec{f} = m\vec{a} = m \frac{d\vec{v}}{dt} = m \frac{d^2\vec{r}}{dt^2}$$

② The motion of a **quantum** particle is described by the **Schrodinger equation**

$$i\hbar \frac{\partial \Psi}{\partial t} = -\frac{\hbar^2}{2m} \nabla^2 \Psi + V\Psi$$

③ The **dynamics and statics of bulk materials** such as fluids and solids are all described by **differential equations**.

In general, we can **classify ordinary differential equations** into **three** major categories:

initial-value problems	time-dependent equations with given initial conditions
boundary-value problems	differential equations with specified boundary conditions
eigenvalue problems	solutions for selected parameters (eigenvalues) in the equations

# Initial-value problems

- Typically, initial-value problems involve **dynamical systems**. For example, the motion of the moon, earth, and sun, the dynamics of a rocket, or the propagation of ocean waves.
- A **dynamical system** can be described by **a set of first-order differential equations**:

$$\frac{d\vec{y}}{dt} = g(\vec{y}, t) \quad \vec{y} = (y_1, y_2, \dots, y_l) \quad \text{the generalized position vector}$$

$$g(\vec{y}, t) = [g_1(\vec{y}, t), g_2(\vec{y}, t), \dots, g_l(\vec{y}, t)] \quad \text{the generalized velocity vector}$$

# Example

- A particle moving in one dimension under an **elastic force**

$$\vec{f} = m\vec{a} = m \frac{d\vec{v}}{dt} = -k\vec{x}$$

- Define  $y_1=x$ ;  $y_2=v$ ;

- Then we obtain:

$$\frac{dy_1}{dt} = y_2,$$
$$\frac{dy_2}{dt} = -\frac{k}{m} y_1,$$

If the **initial position**  $y_1(0)=x(0)$  and the **initial velocity**  $y_2(0) = v(0)$  are given, we can solve the problem numerically.

# The Euler method

$$\frac{dy}{dt} \approx \frac{y_{i+1} - y_i}{t_{i+1} - t_i} \approx g(y_i, t_i)$$

$$y_{i+1} = y_i + \tau g_i + O(\tau^2)$$

$$\tau = t_{i+1} - t_i$$

The **accuracy** of this algorithm is relatively **low**. At the end of the calculation after a total of  $n$  steps, the **error accumulated** in the calculation is on the order of  $nO(\tau^2) \sim O(\tau)$ .

We can formally rewrite the above equation as an integral

$$y_{i+j} = y_i + \int_{t_i}^{t_{i+j}} g(y, t) dt$$

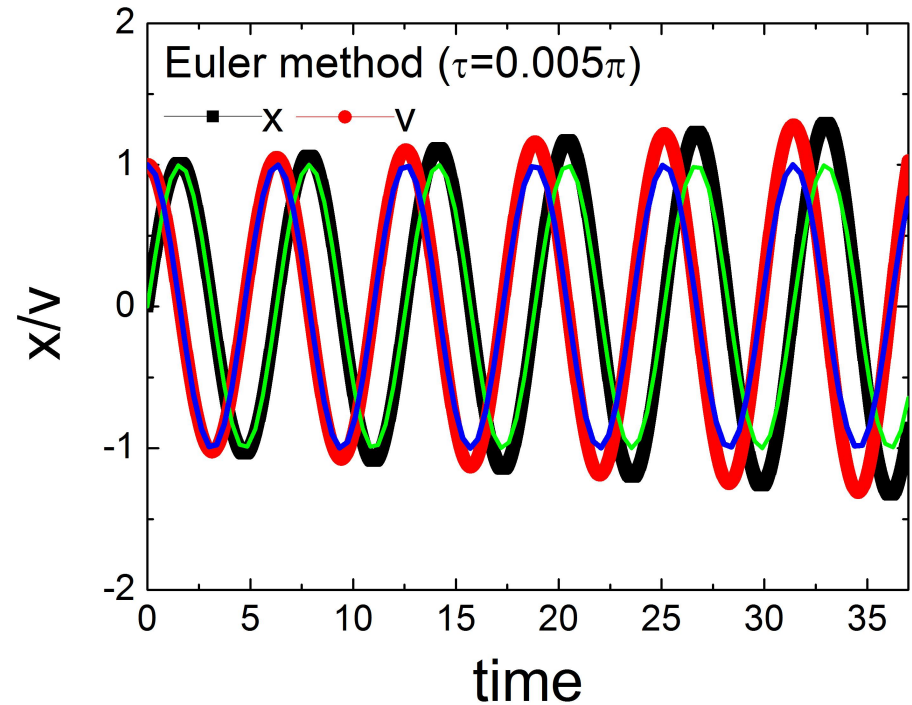
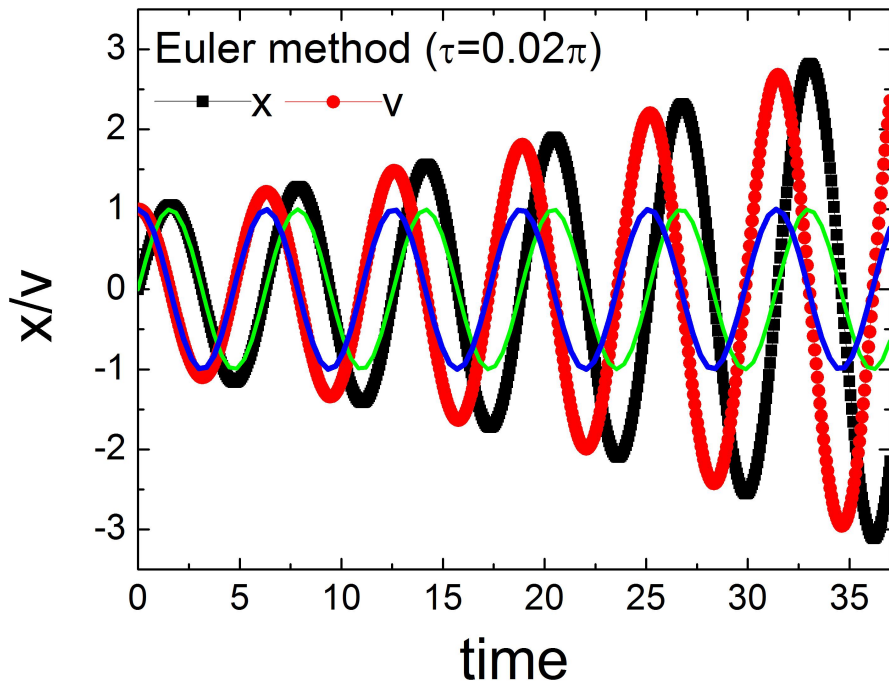
which is the **exact** solution if the **integral** can be obtained **exactly**.

- Because we can not obtain the integral exactly in general, we have to **approximate** it.
- The accuracy in the **approximation of the integral** determines the accuracy of the **solution**.
- If we take the **simplest case** of  $j = 1$  and approximate  $g(y, t) = g_i$  in the integral, we recover the **Euler algorithm**.



# Code example

- [4.1.Euler.cpp](#) (1.3.Intro.cpp)



# Predictor-corrector method

- Use the solution from the **Euler method** as the **starting point**.
- Use a **numerical quadrature** to carry out the integration.
- For example, if we choose  $j = 1$  and use the **trapezoid rule** for the integral.

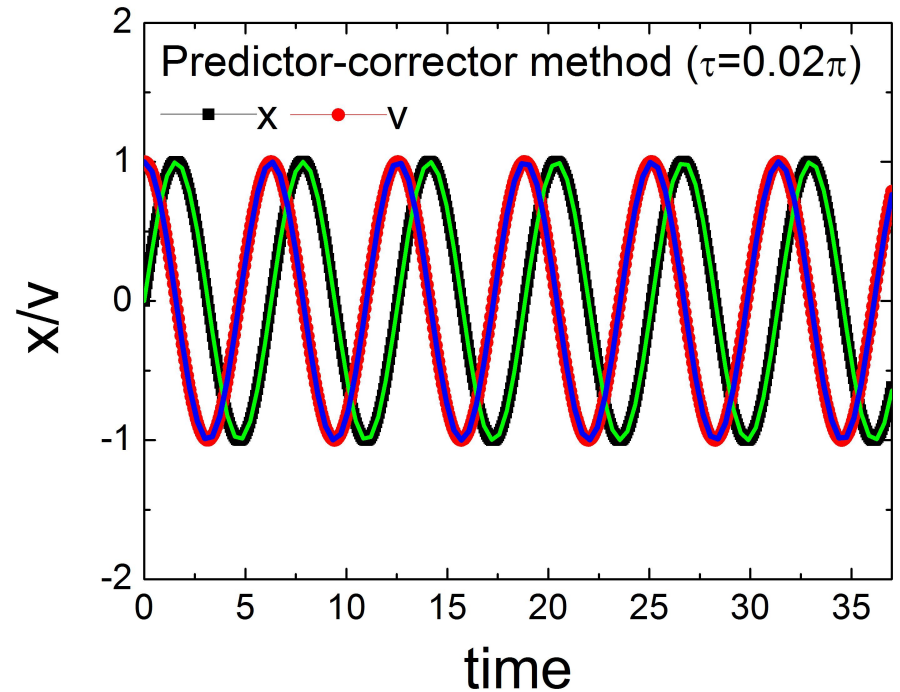
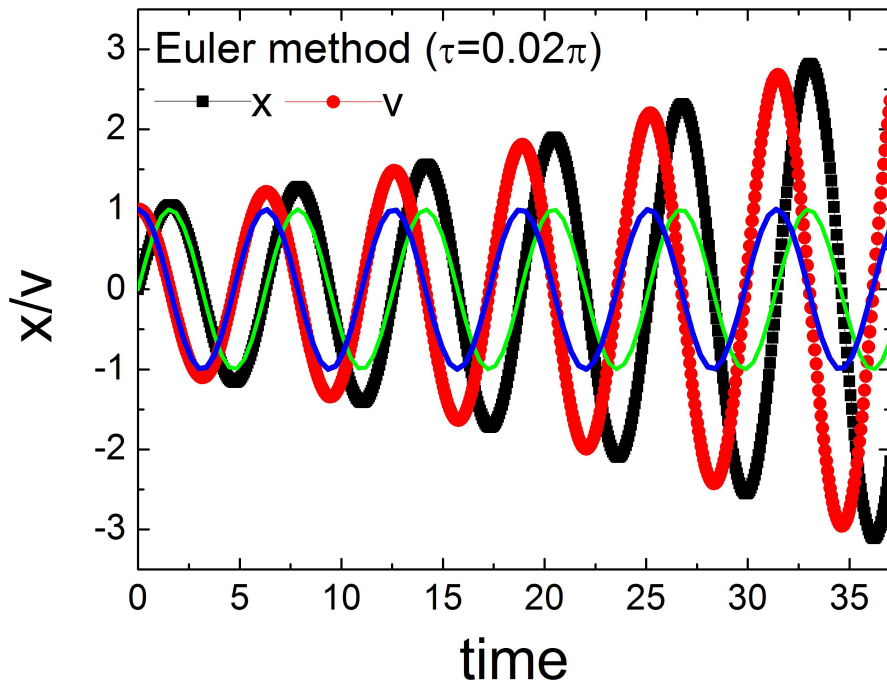
$$y_{i+1} = y_i + \frac{\tau}{2} (g_i + g_{i+1}) + O(\tau^3)$$

# Code example

- **The harmonic oscillation.**
- Euler method: poor accuracy with  $t = 0.02\pi$ .
- Predictor-corrector method: much better?
  
- // Predict the next position and velocity
- $x[i+1] = x[i] + v[i] * dt;$
- $v[i+1] = v[i] - x[i] * dt;$
  
- // Correct the new position and velocity
- $x[i+1] = x[i] + (v[i] + v[i+1]) * dt / 2;$
- $v[i+1] = v[i] - (x[i] + x[i+1]) * dt / 2;$

# Code example

## 4.2. Predictor-Corrector.cpp



- Another way to improve an algorithm is by **increasing the number of mesh points j**. Thus we can apply a **better quadrature** to the integral.

$$y_{i+j} = y_i + \int_{t_i}^{t_{i+j}} g(y, t) dt$$

- For example, take  $j = 2$  and then use the **linear interpolation scheme** to **approximate  $g(y, t)$**  in the integral from  $g_i$  and  $g_{i+1}$ :

$$g(y, t) = \frac{t - t_i}{\tau} g_{i+1} - \frac{(t - t_{i+1})}{\tau} g_i + O(\tau^2)$$

Now if we carry out the integration with  $g(y, t)$  given from this equation, we obtain a new algorithm

$$y_{i+2} = y_i + 2\tau g_{i+1} + O(\tau^3)$$

which has an **accuracy** one order **higher than** that of the **Euler algorithm**.

However, we need the values of the **first two points** in order to start this algorithm, because  $g_{i+1} = g(y_{i+1}, t_{i+1})$ .

We can make the accuracy even higher by using a better quadrature.

For example, we can take  $j = 2$  in above equation and apply the **Simpson rule** to the integral. Then we have

$$y_{i+2} = y_i + \frac{\tau}{3} (g_{i+2} + 4g_{i+1} + g_i) + O(\tau^5)$$

This implicit algorithm can be used as the **corrector** if the previous **algorithm** is used as **the predictor**.

# A car jump over the yellow river

- 1997年，香港回归前夕，柯受良驾驶跑车成功飞越了黄河天堑壶口瀑布，长度达55米。飞越当天刮着大风，第一次飞越没有成功，但第二次成功了，其中有过很多危险的动作，但他都安全度过了，因此获得了“亚洲第一飞人”的称号。



1953-2003



- Let us take a simple model of a car jump over a gap as an example.
- The **air resistance** on a moving object is roughly given by  $f_r = -\kappa v v = -cA\rho v v$ , where  $A$  is cross section of the moving object,  $\rho$  is the density of the air, and  $c$  is a coefficient that accounts for all the other factors.
- So the motion of the system is described by the equation set

$$\frac{d\vec{r}}{dt} = \vec{v}, \quad \frac{d\vec{v}}{dt} = \vec{a} = \frac{\vec{f}}{m},$$

$$\vec{f} = -mg\hat{y} - \kappa v \vec{v}.$$

# Code example

- $f$  is the **total force** on the car of a total mass  $m$ . Here  $y$  is the unit vector pointing upward.
- Assuming that we have the first point given, that is,  $r_0$  and  $v_0$  at  $t = 0$ .

[4.3.FlyingCar.cpp](#)

# The Runge–Kutta method

Formally, we can expand  $y(t+\tau)$  in terms of the quantities at  $t$  with the **Taylor expansion**:

$$y(t + \tau) = y + \tau y' + \frac{\tau^2}{2} y'' + \frac{\tau^3}{3!} y^{(3)} + \dots$$

A particle moving in one dimension under an

**elastic force**  $\vec{f} = m\vec{a} = m \frac{d\vec{v}}{dt} = -k\vec{x}.$

We know the initial condition  $x(0), v(0).$

- $x(t) = x(0) + x'(0)t + x''(0)t^2/2 + \dots$
- $v(t) = v(0) + v'(0)t + v''(0)t^2/2 + \dots$
- $x' = v; \quad x'' = v' = -kx/m$
- $v'' = -kx'/m = -kv/m$
- The same process for higher orders  $x^n$  and  $v^n$ :
- $x''' = v'';$
- $v''' = -kv'/m = k^2x/m^2;$
- $x'''' = v'''';$
- $v'''' = k^2v/m^2$

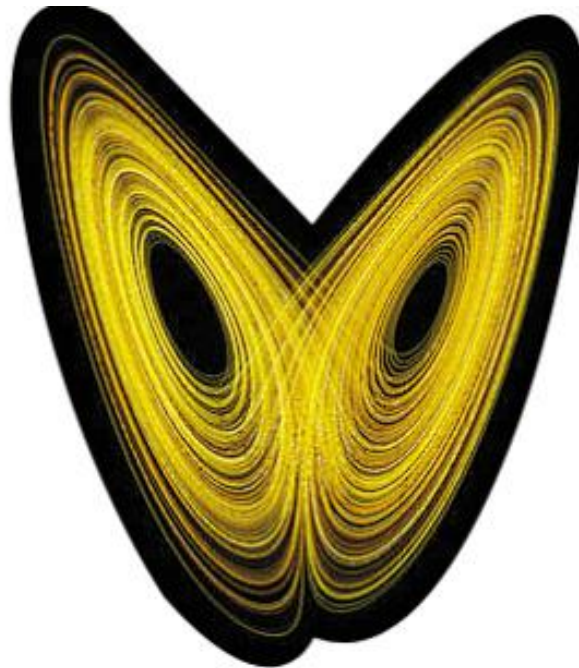
# Code example

4th-order Runge–Kutta algorithm for the  
harmonic oscillator

[4.4.RungeKutta.cpp](#)

# Chaotic dynamics

- nonlinear item
- nonlinear physics
- chaos



# An undergraduate project

PHYSICAL REVIEW B 76, 054414 (2007)

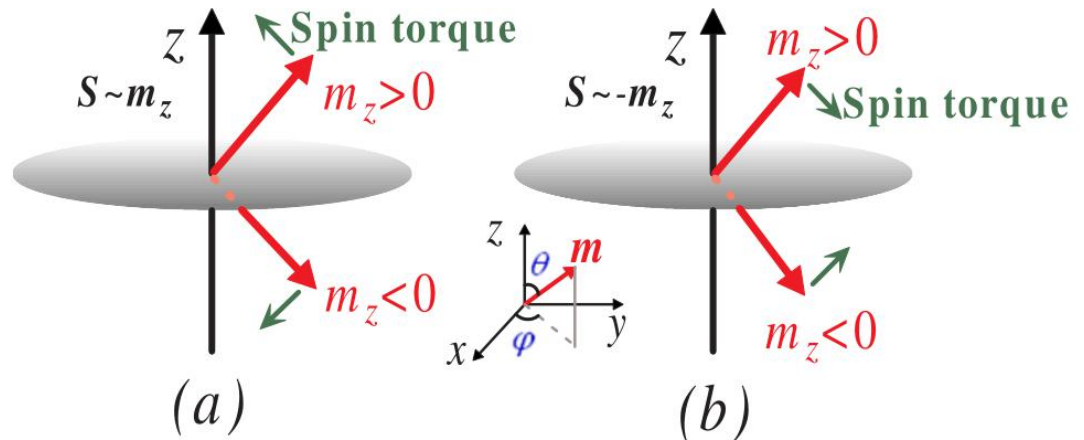
## Magnetization oscillation in a nanomagnet driven by a self-controlled spin-polarized current: Nonlinear stability analysis

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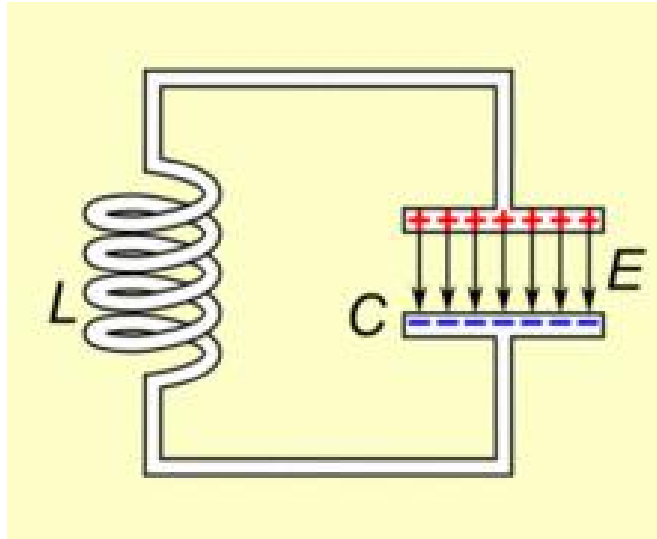
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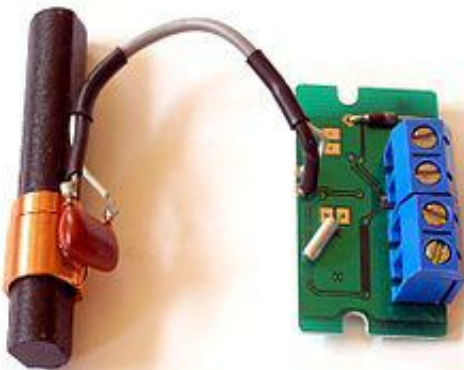


$$\frac{1 + \alpha^2}{\gamma} \frac{dm}{dt} = -m \times h_{eff} - \alpha m \times (m \times h_{eff}) - m \times (m \times S)$$

# An LC circuit



- An LC circuit, also called a **resonant circuit**, tank circuit, or tuned circuit, is an electric circuit consisting of an **inductor**, represented by the letter L, and a **capacitor**, represented by the letter C, connected together. The circuit can act as an electrical resonator, an electrical analogue of a tuning fork, storing energy oscillating at the circuit's resonant frequency.





# Equations of LC circuit

$$V_C = V_L.$$

$$\frac{d^2 i_L(t)}{dt^2} + \frac{1}{LC} i_L(t) = 0.$$

$$i_C = -i_L.$$

$$\omega_0 = \frac{1}{\sqrt{LC}}.$$

$$V_L(t) = L \frac{di_L}{dt}$$

$$\frac{d^2 i_L(t)}{dt^2} + \omega_0^2 i_L(t) = 0.$$

$$i_C(t) = C \frac{dV_C}{dt}.$$

- [https://en.wikipedia.org/wiki/LC\\_circuit](https://en.wikipedia.org/wiki/LC_circuit)
- <https://baike.baidu.com/item/LC%E6%8C%AF%E8%8D%A1%E7%94%B5%E8%B7%AF/2139277?fr=aladdin>

# Homework

- Use the 4th order Runge-Kutta method to solve a LC circuit with resistance & excitation.

# Boundary-value problems

- The solution of the **Poisson equation** with a given charge distribution and known boundary values of the electrostatic potential.
- **Wave equations** with given boundary conditions.
- The **stationary Schrodinger equation** with a given potential and boundary conditions.

# One-dimensional example

$$u'' = f(u, u'; x)$$

- Where  $u$  is a function of  $x$ ,  $u'$  and  $u''$  are the 1st and 2nd derivatives of  $u$  with respect to  $x$ ;  $f(u, u'; x)$  is a function of  $u$ ,  $u'$ , and  $x$ .
- **Either  $u$  or  $u'$  is given at each boundary point.** We can always choose a coordinate system so that the **boundaries** of the system are at  **$x=0$  and  $x=1$**  without losing any generality if the system is finite.

- For example, if the **actual boundaries** are at  $x = x_1$  and  $x = x_2$  for a given problem, we can always bring them back to  $x'=0$  and  $x'=1$  by moving and scaling with a transformation:  $x' = (x - x_1) / (x_2 - x_1)$
- For problems in **one dimension**, we can have a total of **four possible types** of **boundary conditions**:
  - (1)  $u(0) = u_0$  and  $u(1) = u_1$ ;
  - (2)  $u(0) = u_0$  and  $u'(1) = v_1$ ;
  - (3)  $u'(0) = v_0$  and  $u(1) = u_1$ ;
  - (4)  $u'(0) = v_0$  and  $u'(1) = v_1$ .
- **(2) is the same as (3)** by reversing the direction.

- The **boundary-value** problem is **more difficult to solve** than the similar **initial-value problem** with the differential equation.
- For example, **if** we want to solve **an initial-value problem** and the initial conditions  $u(0) = u_0$  **and**  $u'(0) = v_0$ , the solution will follow the algorithms discussed earlier.
- However, for the **boundary-value problem**, we know **only**  $u(0)$  **or**  $u'(0)$ , which is **not sufficient** to start an algorithm for the initial-value problem without some further work.

## Example:

### longitudinal vibrations along an elastic rod

- The equation describing the **stationary solution of elastic waves** is  $u''(x) = -k^2 u(x)$
- If **both ends** ( $x=0$  and  $x=1$ ) of the rod are **fixed**, the boundary conditions are  **$u(0)=u(1)=0$** .
- If one end ( $x=0$ ) is **fixed** and the other end ( $x=1$ ) is **free**, the boundary conditions are  **$u(0)=0$  and  $u'(1)=0$** .

- For example, if **both ends** of the rod are **fixed**, the **eigenfunctions**

$$u_l(x) = \sqrt{2} \sin k_l x$$

are the possible solutions of the differential equation.

- Here the eigenvalues are given by

$$k_l^2 = (l\pi)^2 \quad \text{with } l = 1, 2, \dots, \infty.$$



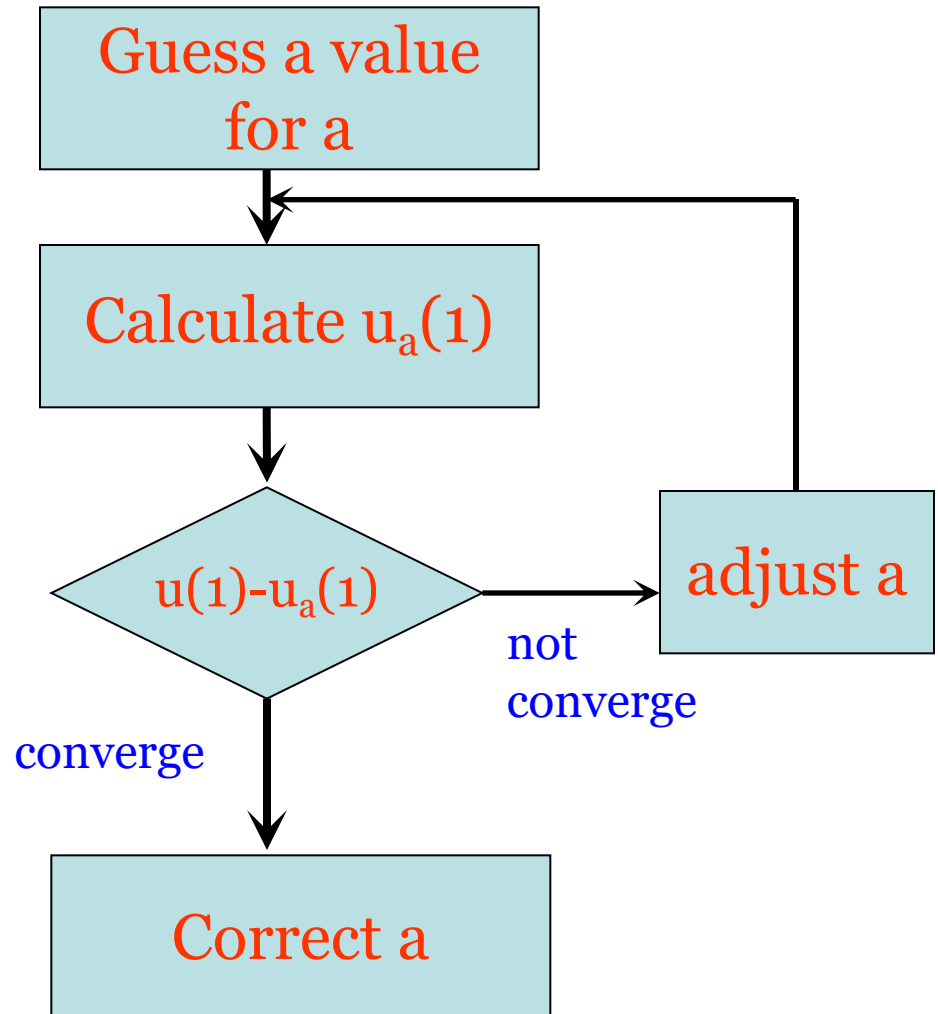
# The shooting method

- The key here is to make the problem **look like an initial-value** problem by **introducing an adjustable parameter**; the solution is then obtained by **varying the parameter**.
- For example, given  $u(0)$  and  $u(1)$ , we can **guess a value of  $u'(0)=a$** , where  $a$  is the parameter to be **adjusted**.

# The shooting method

- For a **specific  $a$** , the value of the function  $u_\alpha(1)$ , resulting from the integration with  $u'(0)=a$  to  $x = 1$ , would **not be the same** as  $u_1$ .
- The idea of the **shooting method** is to use one of the root search algorithms to **find the appropriate  $\alpha$**  that ensures  $f(\alpha)=u_\alpha(1)-u(1)=0$  within a given **tolerance  $\delta$** .

# The shooting method



# Example

$$u'' = -\frac{\pi^2}{4}(u + 1)$$

- With given boundary conditions  $u(0) = 0$  and  $u(1) = 1$ , We can **define new variables**  $y_1 = u$  and  $y_2 = u'$ ;

$$\frac{dy_1}{dx} = y_2, \quad \frac{dy_2}{dx} = -\frac{\pi^2}{4}(y_1 + 1)$$

- Assume that this equation set has the **initial values**  $y_1(0) = 0$  and  $y_2(0) = \alpha$ .
- Here  $\alpha$  is a parameter to be **adjusted** in order to have  $f(\alpha) = u_\alpha(1) - 1 = 0$ .
- We can combine the **secant method for the root search** and the **4th-order Runge–Kutta method for initial-value problems** to solve the above equation set.

# Linear equations

- Many eigenvalue or boundary-value problems are in the form of **linear equations**, such as

$$u'' + d(x)u' + q(x)u = s(x)$$

- Assume that the boundary conditions are  $u(0) = u_0$  and  $u(1) = u_1$ . If all  $d(x)$ ,  $q(x)$ , and  $s(x)$  are **smooth**, we can solve the equation with the **shooting method** as shown above.
- However, an extensive search for the parameter  $\alpha$  from  $f(\alpha) = u_\alpha(1) - u_1 = 0$  is **unnecessary** in this case, because of the **principle of superposition** of linear equations: **any linear combination of the solutions is also a solution of the equation.**

- We need only **two trial solutions**  $u_{\alpha_0}(x)$  and  $u_{\alpha_1}(x)$ , where  $\alpha_0$  and  $\alpha_1$  are two different parameters.
- The correct solution of the equation is given by

$$u(x) = au_{\alpha_0}(x) + bu_{\alpha_1}(x)$$

where  $a$  and  $b$  are determined from  $u(0) = u_0$  and  $u(1) = u_1$ . Note that  $u_{\alpha_0}(0) = u_{\alpha_1}(0) = u(0) = u_0$ .  
So we have

$$a + b = 1,$$

$$u_{\alpha_0}(1)a + u_{\alpha_1}(1)b = u_1,$$

# Code example

- [4.5.Boundary.cpp](#)

$$u'' = -\frac{\pi^2}{4}(u + 1)$$

