



Computational Physics

Physics problems and Solutions

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Computational Physics 2020-21 (Phys Dep IST, Lisbon)

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Example: system of 1st-order ODEs

- ✓ Solve numerically the following system:

$$\frac{dw}{dx} = \sin(x) + y$$

$$\frac{dy}{dx} = \cos(x) - w$$

- ✓ Initial values:

$$w(0) = 0$$

$$y(0) = 0$$

- ✓ variables and functions:

$$\begin{cases} y^{(0)} = w & f^{(0)} = \sin(x) + y^{(1)} \\ y^{(1)} = y & f^{(1)} = \cos(x) - y^{(0)} \end{cases}$$

- ✓ runge-kutta 4th-order iterations:

$$\begin{cases} K_1^{(0)} = hf^{(0)}(x_n, y_n^{(0)}, y_n^{(1)}) \\ K_1^{(1)} = hf^{(1)}(x_n, y_n^{(0)}, y_n^{(1)}) \\ K_2^{(0)} = hf^{(0)}\left(x_n + \frac{h}{2}, y_n^{(0)} + \frac{K_1^{(0)}}{2}, y_n^{(1)} + \frac{K_1^{(1)}}{2}\right) \\ K_2^{(1)} = hf^{(1)}\left(x_n + \frac{h}{2}, y_n^{(0)} + \frac{K_1^{(0)}}{2}, y_n^{(1)} + \frac{K_1^{(1)}}{2}\right) \\ K_3^{(0)} = hf^{(0)}\left(x_n + \frac{h}{2}, y_n^{(0)} + \frac{K_2^{(0)}}{2}, y_n^{(1)} + \frac{K_2^{(1)}}{2}\right) \\ K_3^{(1)} = hf^{(1)}\left(x_n + \frac{h}{2}, y_n^{(0)} + \frac{K_2^{(0)}}{2}, y_n^{(1)} + \frac{K_2^{(1)}}{2}\right) \\ K_4^{(0)} = hf^{(0)}\left(x_n + h, y_n^{(0)} + K_3^{(0)}, y_n^{(1)} + K_3^{(1)}\right) \\ K_4^{(1)} = hf^{(1)}\left(x_n + h, y_n^{(0)} + K_3^{(0)}, y_n^{(1)} + K_3^{(1)}\right) \end{cases}$$

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2nd order ODE: numerical solutions

✓ Taylor method (2nd order) - Stormer-Verlet

using the numerical approximation for the 2nd-order derivative:

$$\left. \frac{d^2 y}{dt^2} \right|_n \simeq \frac{y_{n+1} - 2y_n + y_{n-1}}{(\delta t)^2} + O[(\delta t)^2]$$

the differential equation becomes:

$$\left. \frac{d^2 y}{dt^2} \right|_n = f[t_n, y(t_n)] \Rightarrow y_{n+1} = -y_{n-1} + 2y_n + (\delta t)^2 f[t_n, y(t_n)] + O[(\delta t)^4]$$

✓ algorithm

time: $\delta t = (t_f - t_0)/n$

initial conditions: $t(0) \equiv t_0$

$y(0) \equiv y_0$

$\left. \frac{dy}{dt} \right|_{t=0} \equiv \left(\frac{dy}{dt} \right)_0$

first iteration (n=0): $y_1 = 2y_0 - y_{-1} + (\delta t)^2 f[t_n, y(t_n)]$

$y_{-1} \equiv y(x-h) \simeq y(x) - h \left. \frac{dy}{dt} \right|_x + \frac{h^2}{2} \left. \frac{d^2 y}{dt^2} \right|_x = y_0 - h v_0 + \frac{h^2}{2} f[t_0, y(t_0)]$

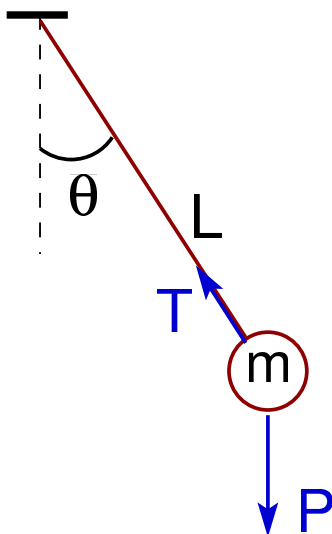
following iterations (n=1,...): $y_{n+1} = -y_{n-1} + 2y_n + (\delta t)^2 f(t_n, y_n)$

$t_{n+1} = t_0 + (n+1)\delta t$



simple pendulum

pendulum motion



Lagrangian: $\mathcal{L} = \frac{1}{2} m \ell^2 \dot{\theta}^2 + mg\ell \cos \theta$

eq. of motion: $\frac{d^2 \theta}{dt^2} + \frac{g}{\ell} \sin \theta = 0$

initial conditions: $\theta(0) = \theta_0 \quad \dot{\theta}(0) = \omega_0$

simplify equation

characteristic time: $t_c = \sqrt{\frac{\ell}{g}}$

variable change: $\tau = \frac{t}{t_c}$

$$\frac{d^2 \theta}{d\tau^2} + \sin \theta = 0$$

reduce to a system of 1st-order diff equations

$$\begin{cases} \frac{d\theta}{dt} = \omega \\ \frac{d\omega}{dt} = -\sin(\theta) \end{cases} \Rightarrow \begin{cases} \frac{dx_1}{dt} = x_2 = f_1(t, x_1, x_2) \\ \frac{dx_2}{dt} = -\sin(x_1) = f_2(t, x_1, x_2) \end{cases}$$

Solution methods:

- Euler
- Euler-Cromer
- Euler-Verlet
- Runge-Kutta



simple pendulum: solutions

Euler method

velocity is computed at the beginning of interval $(t, t + \Delta t)$

$$\begin{cases} x_{2,n+1} = x_{2,n} + (\Delta t) f_2(x_{1,n}) \\ x_{1,n+1} = x_{1,n} + (\Delta t) f_1(x_{2,n}) \end{cases}$$

$$\omega_{n+1} = \omega_n + (\Delta t) [-\sin \theta_n]$$

$$\theta_{n+1} = \theta_n + (\Delta t) [\omega_n]$$

Euler-Cromer method

coordinate uses velocity computed at the end of the interval $(t, t + \Delta t)$ improving behavior

$$\begin{cases} x_{2,n+1} = x_{2,n} + (\Delta t) f_2(x_{1,n}) \\ x_{1,n+1} = x_{1,n} + (\Delta t) f_1(x_{2,n+1}) \end{cases}$$

$$\omega_{n+1} = \omega_n + (\Delta t) [-\sin \theta_n]$$

$$\theta_{n+1} = \theta_n + (\Delta t) [\omega_{n+1}]$$

Euler-Verlet method

it uses the 2nd-derivative operator $\ddot{y} = (\Delta t)^{-2} (y_{n+1} - 2y_n + y_{n-1}) + O[(\Delta t)^2]$

$$\begin{cases} \theta_{n+1} = 2\theta_n - \theta_{n-1} + (\Delta t)^2 f(\theta_n) \\ \omega_n = \frac{\theta_{n+1} - \theta_{n-1}}{2(\Delta t)} \end{cases}$$

first iteration $\begin{cases} \theta_1 = \theta_0 - (\Delta t) \omega_0 + \frac{(\Delta t)^2}{2} f(\theta_0) \\ \omega_0 \end{cases}$

next iterations $\begin{cases} \theta_2 = 2\theta_1 - \theta_0 + (\Delta t)^2 f(\theta_1) \\ \omega_1 = \frac{\theta_2 - \theta_0}{2(\Delta t)} \end{cases}$



simple pendulum: solutions (cont.)

Runge-Kutta method (4th order)

$$K_{11} = f_1(t_n, x_{1,n}, x_{2,n})$$

$$K_{12} = f_2(t_n, x_{1,n}, x_{2,n})$$

$$K_{21} = f_1\left(t_n + \frac{h}{2}, x_{1,n} + \frac{h}{2}K_{11}, x_{2,n} + \frac{h}{2}K_{12}\right)$$

$$K_{22} = f_2\left(t_n + \frac{h}{2}, x_{1,n} + \frac{h}{2}K_{11}, x_{2,n} + \frac{h}{2}K_{12}\right)$$

$$K_{31} = f_1\left(t_n + \frac{h}{2}, x_{1,n} + \frac{h}{2}K_{21}, x_{2,n} + \frac{h}{2}K_{22}\right)$$

$$K_{32} = f_2\left(t_n + \frac{h}{2}, x_{1,n} + \frac{h}{2}K_{21}, x_{2,n} + \frac{h}{2}K_{22}\right)$$

$$K_{41} = f_1(t_n + h, x_{1,n} + hK_{31}, x_{2,n} + hK_{32})$$

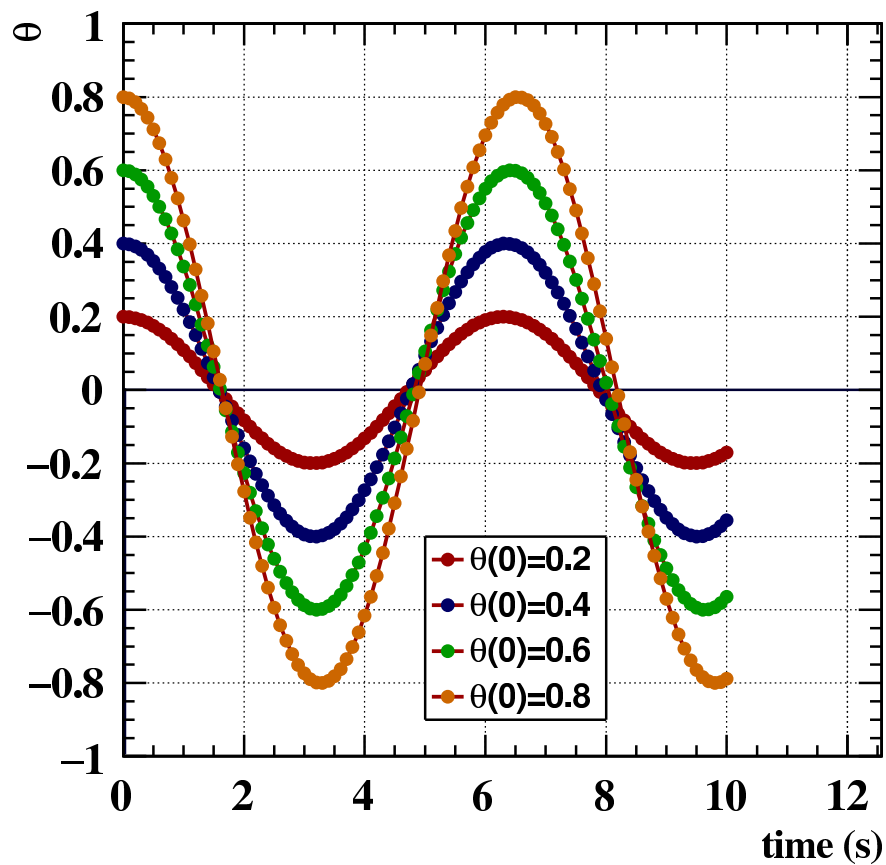
$$K_{42} = f_2(t_n + h, x_{1,n} + hK_{31}, x_{2,n} + hK_{32})$$

$$x_{1,n+1} = x_{1,n} + \frac{h}{6} (K_{11} + 2K_{21} + 2K_{31} + K_{41})$$

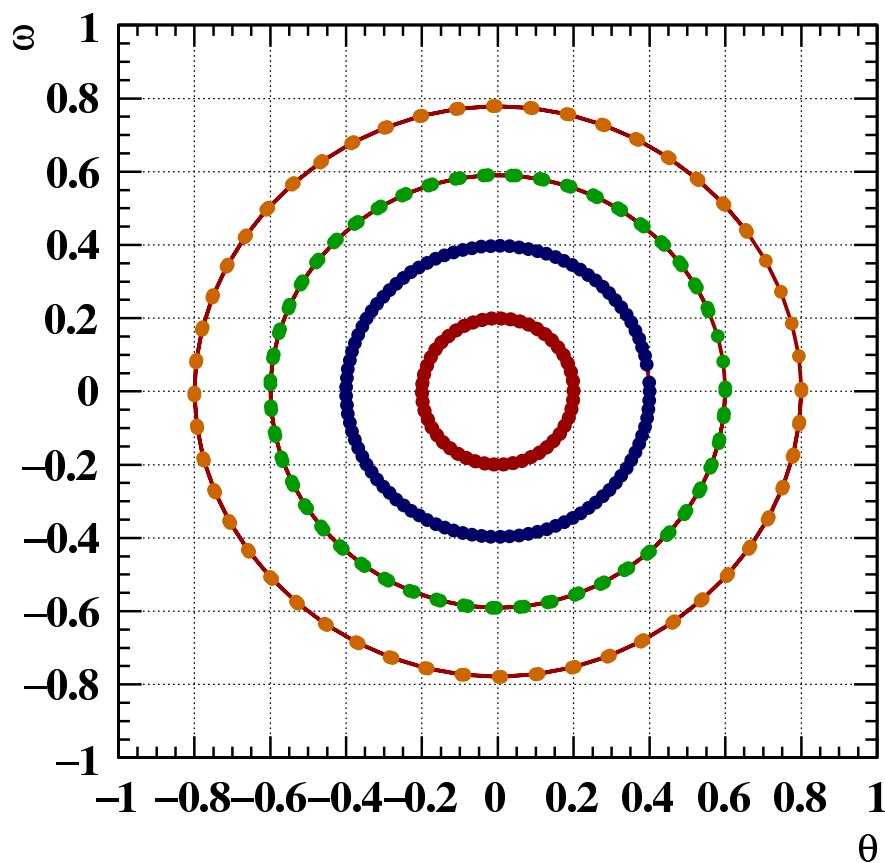
$$x_{2,n+1} = x_{2,n} + \frac{h}{6} (K_{12} + 2K_{22} + 2K_{32} + K_{42})$$



simple pendulum: RK4 solution

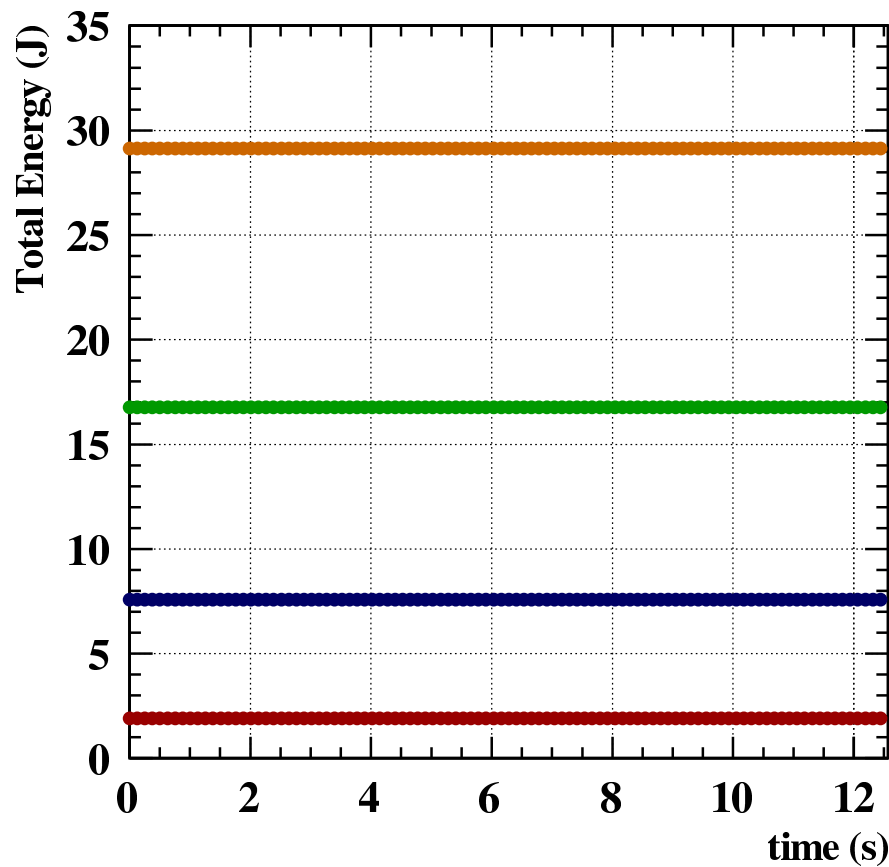


simple pendulum: RK4 solution

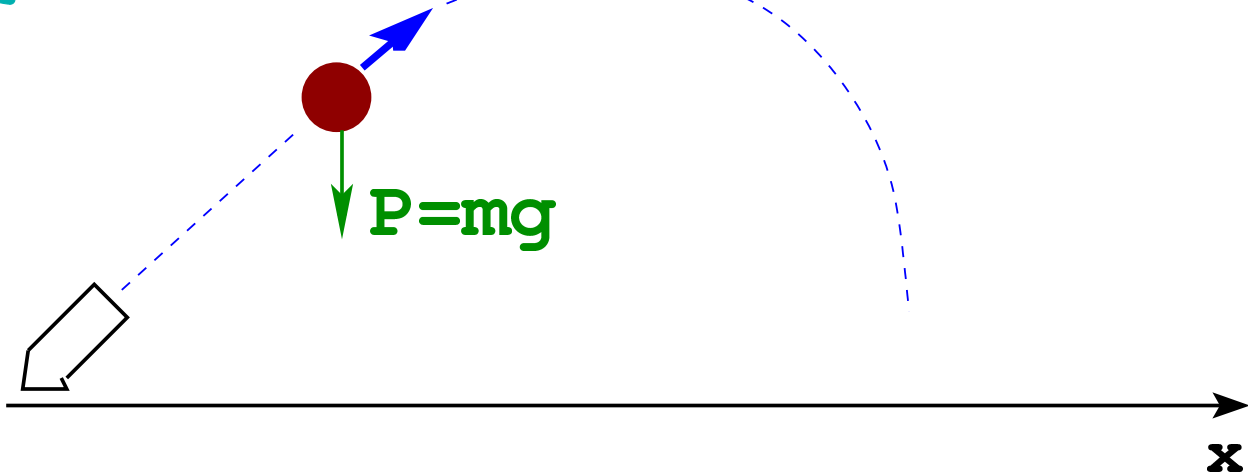




simple pendulum: RK4 solution



projectile motion



✓ Forces:

- ▶ gravitic: $\vec{P} = m \vec{g}$
- ▶ friction: $\vec{F}_a = c v^2 \left(\frac{\vec{v}}{v} \right)$

✓ Initial conditions

- ▶ position: $(x, y) = (0, 0)$
- ▶ velocity: \vec{v}_0



projectile motion: multi-dimensions

- ✓ equation of motion

$$\frac{d^2 \vec{r}}{dt^2} = -mg \vec{e}_y$$

- ✓ in terms of coordinates

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

$$\frac{d^2 y}{dt^2} = -mg$$

- ✓ system of 1st-order differential equations

$$\frac{dx}{dt} = v_x$$

$$\frac{dy}{dt} = v_y$$

$$\frac{dv_x}{dt} = 0$$

$$\frac{dv_y}{dt} = -mg$$

renaming variables

$$x, y, v_x, v_y \rightarrow x_1, x_2, x_3, x_4$$

$$\frac{dx_1}{dt} = x_3 = f_1(t, x_1, x_2, x_3, x_4)$$

$$\frac{dx_2}{dt} = x_4 = f_2(t, x_1, x_2, x_3, x_4)$$

$$\frac{dx_3}{dt} = 0 = f_3(t, x_1, x_2, x_3, x_4)$$

$$\frac{dx_4}{dt} = -mg = f_4(t, x_1, x_2, x_3, x_4)$$

ODEsolver

- Runge-Kutta
- Euler
- Euler-Verlet

needs

- number variables
- functions

[Library?](#)

Physics Problem

eqs of motion



[user problem](#)



ODEsolver class

ODEsolver class

constructor

```
ODEsolver(const vector<TF1>&);
```

methods

```
void SetODEfunctions(const vector<TF1>&);
```

```
vector<ODEpoint> RK4(ODEpoint i,  
                     double step,  
                     double time);
```



ODEpoint class

ODEpoint class

```
class ODEpoint {
public:
    ODEpoint(int fndim = 0, const double* x = nullptr);
    ODEpoint(const ODEpoint&);
    ~ODEpoint();

    int Ndim() const {return ndim;};
    double T() const {return x[0];};
    double X(int i) const {return x[i+1];};
    double* GetArray() { return x;}

    ODEpoint operator*(double) const;
    ODEpoint operator+(const ODEpoint&) const;
    ODEpoint operator-(const ODEpoint&) const;

    void operator=(const ODEpoint&);

    const double& operator[] (int) const;
    double& operator[] (int);

private:
    int ndim; //nb of dependent variables
    double* x; //independent variable + dependent variables (ndim+1)
};
```



ODEsolver class

ODEpoint class

```
class ODEsolver {
public:
    ODEsolver(const vector<TF1>&);
    ~ODEsolver();

    SetODEfunc(const vector<TF1>&);

    vector<ODEpoint> Euler(ODEpoint i, double step, double T);
    vector<ODEpoint> PredictorCorrector(ODEpoint i, double step, double T);
    vector<ODEpoint> LeapFrog(ODEpoint i, double step, double T);

    vector<ODEpoint> RK2(ODEpoint i, double step, double T);
    vector<ODEpoint> RK4(ODEpoint i, double step, double T);

    (...)

private:
    vector<TF1> F;
    (...)
};
```



Runge-Kutta: adaptive step

- ✓ How to find the most appropriate step size?
 - ▶ a step h too large leads to a big truncation error
 - ▶ a step h too small leads to waste of computer resources
- ✓ We need an adaptive method to have an estimate of the truncation error and adjust the step size
 - ▶ two embedded integration formulas of order m and $m + 1$ will be used to evaluate error

$$E^{(i)}(h) = y_{m+1}^{(i)}(x+h) - y_m^{(i)}(x+h)$$

Runge-Kutta-Fehlberg formulas of orders 5 and 4

a set of 1st-order equations to solve
 $\frac{dy^{(i)}}{dt} = f^{(i)}(t, y^{(i)})$ [i dependent variables]

$$K_1^{(i)} = h f^{(i)}(x, y^{(i)})$$

$$K_{u+1}^{(i)} = h f^{(i)}\left(x + h A_u, y^{(i)} + \sum_{\ell=0}^{u-1} B_{u\ell} K_{\ell+1}^{(i)}\right)$$

$$[u = 1, 2, 3, 4, 5]$$

$$y_5^{(i)}(x+h) = y^{(i)}(x) + \sum_{u=0}^5 C_u K_{u+1}^{(i)} \quad (5\text{th-order})$$

$$y_4^{(i)}(x+h) = y^{(i)}(x) + \sum_{u=0}^5 D_u K_{u+1}^{(i)} \quad (4\text{th-order})$$



Cash and Karp coefficients

$$E^{(i)}(h) = y_5^{(i)}(x+h) - y_4^{(i)}(x+h) = \sum_{u=0}^5 (C_u - D_u) K_{u+1}^{(i)}$$

errors on dependent variables $y^{(i)}$

error estimate

u	A_u	$B_{u\ell}$					C_u	D_u
0	-	-	-	-	-	-	$\frac{37}{378}$	$\frac{2825}{27648}$
1	$\frac{1}{5}$	$\frac{1}{5}$	-	-	-	-	0	0
2	$\frac{3}{10}$	$\frac{3}{40}$	$\frac{9}{40}$	-	-	-	$\frac{250}{621}$	$\frac{18575}{48384}$
3	$\frac{3}{5}$	$\frac{3}{10}$	$-\frac{9}{10}$	$\frac{6}{5}$	-	-	$\frac{125}{594}$	$\frac{13525}{55296}$
4	1	$-\frac{11}{54}$	$\frac{5}{2}$	$-\frac{70}{27}$	$\frac{35}{27}$	-	0	$\frac{277}{14336}$
5	$\frac{7}{8}$	$\frac{1631}{55296}$	$\frac{175}{512}$	$\frac{575}{13824}$	$\frac{44275}{110592}$	$\frac{253}{4096}$	$\frac{512}{1771}$	$\frac{1}{4}$



adaptive step

- ✓ suppose a system of k 1st-order equations to solve corresponding to k dependent variables $\mathbf{y}^{(i)}$
- ✓ an error estimate at step \mathbf{h} can be provided by,

$$\varepsilon(h) = \sqrt{\frac{1}{k} \sum_{i=0}^{k-1} E_i^2(h)}$$

- ✓ Given that the error corresponds to the fourth-order formula, the ratio between errors with different step sizes,

$$\frac{\varepsilon(h_1)}{\varepsilon(h_2)} = \left(\frac{h_1}{h_2}\right)^5$$

- ✓ Last step \mathbf{h}_n provided us with an error estimate of $\varepsilon(h_n)$
- ✓ an envisaged error tolerance of δ is required
for instance $\delta = 10^{-6}$
- ✓ For reaching that tolerance error
 $\varepsilon(h_{n+1}) = \delta$
we need for the next step a size of,

$$h_{n+1} = \alpha h_n \left[\frac{\delta}{\varepsilon(h_n)} \right]^{1/5}$$

α , safety margin (0.9)

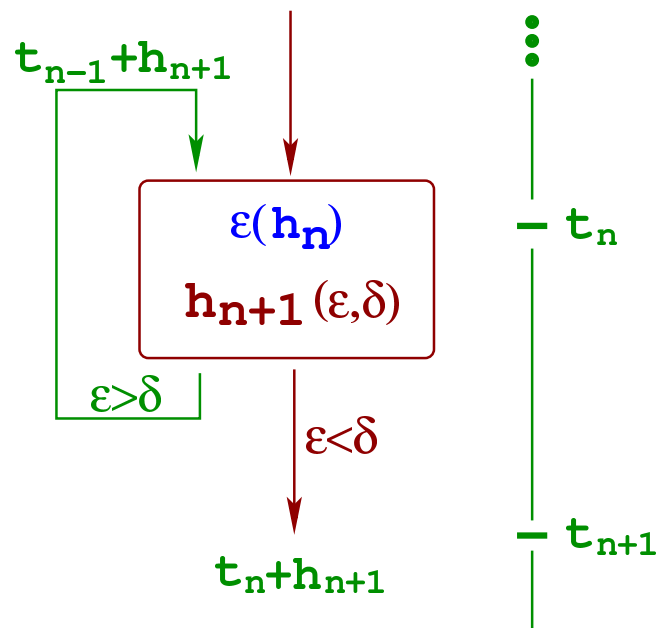


adaptive step algorithm

- ✓ At a given grid point, iterate dependent variables $\mathbf{y}^{(i)}$ using 4th and 5th order Runge-Kutta-Fehlberg algorithms
Compute the error on current iteration by using the 5th order as a reference

$$\varepsilon(h) = \sqrt{\frac{1}{m} \sum_{i=0}^{m-1} E_i^2(h)}$$

- ✓ Compute next step size taking into account error tolerance
 h_0 has to be provided at beginning
1. If the error estimate $\varepsilon(h)$ exceeds the prescribed tolerance δ ,
 - a smaller step size h_{n+1} shall be assigned,
 - a new grid point $t_n = t_{n-1} + h_{n+1}$ computed and
 - the preceding iteration repeated
 2. If the error estimate $\varepsilon(h)$ is smaller than the prescribed tolerance δ , the next grid point $t_{n+1} = t_n + h_{n+1}$ is defined and a new iteration made





adimensional equations

- ✓ In a differential equation involving observables like masses, velocities and time for instance, the arbitrary choice of the units can cause problems for numerical calculations
- ✓ We shall write the equations invariants to unit changes, i.e. written as function of adimensional observables

find the characteristic scales of the problem like for length L_c and velocity V_c and define the reduced variables:

$$v = V_c v_0 \quad (v_0, \text{adimensional velocity})$$

$$\ell = L_c \ell_0$$

- ✓ radioactive decay example:

$$dN = -p \, dt \, N = -\frac{dt}{\tau} N \Rightarrow \frac{dN}{dt} = -\frac{N}{\tau}$$

logistic population model:

$$dN = +\kappa \, dt \, N - \frac{\kappa dt}{a^2} N^2 \Rightarrow \frac{dN}{dt} = +\kappa N - \frac{\kappa}{a^2} N^2$$

ODE's : numerical solution optimization

Nondimensionalization

Decay equation

Radioactive element: τ , decay time

$$\frac{dN}{dt} = -\frac{N}{\tau}$$

$$\frac{N_c}{t_c} \frac{d\hat{N}}{d\hat{t}} = -\frac{N_c \hat{N}}{\tau}$$

$$\frac{\tau}{t_c} \frac{d\hat{N}}{d\hat{t}} = -\hat{N}$$

scaling variable: $t = t_c \hat{t}$
 $N = N_c \hat{N}$
 var. scales $\xrightarrow{\text{dimensional}}$

initial condition: $N(0) = N_0$
 $\hat{N}(0) = \frac{N_0}{N_c}$

Scaling equations:

$$\pi_1 = \frac{\tau}{t_c} = 1$$

$$\pi_2 = \frac{N_0}{N_c} = 1$$

$$\boxed{t_c = \tau \quad N_c = N_0} \quad \text{scales}$$

Numerical equation to solve:

$$\boxed{\frac{d\hat{N}}{d\hat{t}} = -\hat{N}}$$

$$\hat{t}_n = \hat{t}_0 + (n+1) \Delta \hat{t}$$

$$t_n = t_c \hat{t}_n$$

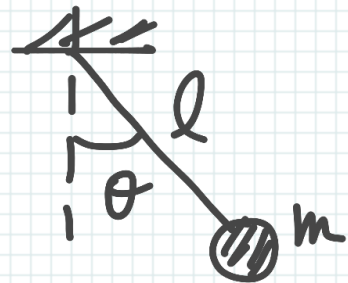
$$\hat{N}_m$$

$$N_m = N_c \hat{N}_m$$

F. Barrow

Pendulum

$$\boxed{\frac{d^2\theta}{dt^2} = -\frac{g}{l} \sin\theta}$$



scaling variables:

$$\begin{cases} t = t_c \hat{t} \\ \theta = \theta_c \hat{\theta} \end{cases} \Rightarrow \frac{\theta_c}{t_c^2} \frac{d^2\hat{\theta}}{d\hat{t}^2} = -\frac{g}{l} \sin(\theta_c \hat{\theta})$$

$$\Rightarrow \left[\frac{l \theta_c}{g t_c^2} \right] \frac{d^2\hat{\theta}}{d\hat{t}^2} = -\sin(\theta_c \hat{\theta})$$

initial conditions:

$$\theta(0) = \theta_0 \Rightarrow \hat{\theta}(0) = \frac{\theta_0}{\theta_c}$$

$$\boxed{\begin{aligned} \theta_c &= \theta_0 \\ t_c &= \sqrt{\theta_0 \frac{l}{g}} \end{aligned}}$$

characteristic time

Numerical equation
to solve :

$$\frac{d^2 \hat{\theta}}{d \hat{t}^2} = -\sin(\theta_0 \hat{\theta})$$

velocity: $\omega = \frac{d\theta}{dt} = \frac{\theta_c}{t_c} \frac{d\hat{\theta}}{d\hat{t}}$

$$\omega_m = \sqrt{\theta_0 g / \ell} \hat{\omega}_m$$

$$\hat{t}_n = \hat{t}_0 + (n+1) \Delta \hat{t}$$

$$t_n = \sqrt{\theta_0 \frac{\ell}{g}} \hat{t}_n$$

$$\hat{\theta}_n$$

$$\theta_n = \theta_0 \hat{\theta}_n$$