$$v_{n+1/2} = v_{n-1/2} + \Delta_t a_n$$
 initial values
$$v_{-1/2} = v_0 - \frac{1}{2} \Delta_t a_0$$

$$x_{n+1} = x_n + \Delta_t v_{n+1/2}$$

$$v_{(-1/2)} = v_0 - \frac{1}{2} \Delta_t a_0$$

$$v_{(-1/2)} = v_0 - \frac{1}{2} \Delta_t a_0$$

What is the step error here?

- we expanded x to 2nd order. Is the step error 3rd order?

Let's do a different derivation to cubic order

- forward and backward x steps:

$$x_{n+1} = x_n + \Delta_t v_n + \frac{1}{2} \Delta_t^2 a_n + \frac{1}{6} \Delta_t^3 \dot{a}_n + O(\Delta_t^4)$$

$$x_{n-1} = x_n - \Delta_t v_n + \frac{1}{2} \Delta_t^2 a_n - \frac{1}{6} \Delta_t^3 \dot{a}_n + O(\Delta_t^4)$$
 add these \rightarrow

$$x_{n+1} = 2x_n - x_{n-1} + \Delta_t^2 a_n + O(\Delta_t^4)$$
consider $x_n - x_{n-1} = x(t_{n-1/2} + \Delta_t/2) - x(t_{n-1/2} - \Delta_t/2)$

$$= [x_{n-1/2} + (\Delta_t/2)v_{n-1/2} + \frac{1}{2}(\Delta_t/2)^2 a_{n-1/2} + \frac{1}{6}(\Delta_t/2)^3 \dot{a}_{n-1/2} + \dots]$$

$$- [x_{n-1/2} - (\Delta_t/2)v_{n-1/2} + \frac{1}{2}(\Delta_t/2)^2 a_{n-1/2} - \frac{1}{6}(\Delta_t/2)^3 \dot{a}_{n-1/2} + \dots]$$

$$= \Delta_t v_{n-1/2} + O(\Delta_t^3) \rightarrow x_n - x_{n-1} = v_{n-1/2} \Delta_t + O(\Delta_t^3)$$
use this in $x_{n+1} = 2x_n - x_{n-1} + \Delta_t^2 a_n + O(\Delta_t^4)$

$$x_{n+1} = 2x_n - x_{n-1} + \Delta_t^2 a_n + O(\Delta_t^4)$$

$$x_n - x_{n-1} = v_{n-1/2} \Delta_t + O(\Delta_t^3) \rightarrow v_{n-1/2} = \frac{1}{\Delta_t} (x_n - x_{n-1}) + O(\Delta_t^2)$$

$$x_{n+1} = x_n + \Delta_t (v_{n-1/2} + \Delta_t a_n)$$

Here it looks like the error in x_{n+1} should be $O(\Delta_t^3)$

However, earlier we saw that

$$v_{n+1/2} = v_n + (\Delta_t/2)a_n + (\Delta_t/2)^2 \dot{a}_n + O(\Delta_t^3)$$

$$v_{n+1/2} = v_{n-1/2} + a_n \Delta_t + O(\Delta_t^3)$$

$$v_{n-1/2} = v_n - (\Delta_t/2)a_n + (\Delta_t/2)^2 \dot{a}_n - O(\Delta_t^3)$$

and so the error in x is actually $O(\Delta_t^4)$

We arrive at an algorithm identical to the Leapfrog algorithm

$$v_{n+1/2} = v_{n-1/2} + \Delta_t a_n$$

$$x_{n+1} = x_n + \Delta_t v_{n+1/2}$$

$$x_{n+1} - x_n = x_n - x_{n-1} + \Delta_t^2 a_n + O(\Delta_t^4)$$

The Leapfrog step error is actually $O(\Delta_t^4)$

The form without explicit v is called the Verlet algorithm

$$x_{n+1} = 2x_n - x_{n-1} + \Delta_t^2 a_n + O(\Delta_t^4)$$

Initial conditions: x₀, x₁

If needed, the velocity obtained as
$$v_{n+1/2} = \frac{1}{\Delta_t}(x_{n+1} - x_n) + O(\Delta_t^2)$$

$$v_{n+1/2} = v_{n-1/2} + \Delta_t a_n$$
$$x_{n+1} = x_n + \Delta_t v_{n+1/2}$$

Julia Leapfrog implementation

```
for i=1:nt
    t=dt*(i-1)
    a=acc(x,t)
    v=v+dt*a
    x=x+dt*v
end
```

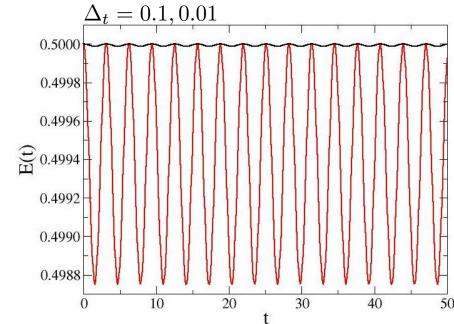
Properties of the method

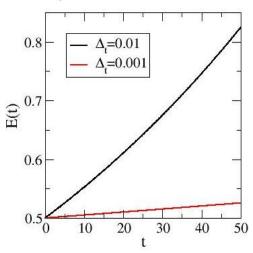
- time-reversal symmetric
- errors bounded for periodic motion
- small step error at low effort

Test: same oscillator as before

Note: Implementation almost identical to Euler

- just swap two lines!





Accumulated errors in the Leapfrog/Verlet algorithm

The number of time steps N can be large; $N=T/\Delta_t$

How does the error at time T depend on T and Δ_t ?

Write the position at time step n as the exact value plus a deviation

$$x_n = x_n^{\rm ex} + \delta_n$$

and use in the Verlet algorithm

$$x_{n+1} = 2x_n - x_{n-1} + \Delta_t^2 a_n + O(\Delta_t^4) \rightarrow$$

$$x_{n+1}^{\text{ex}} - 2x_n^{\text{ex}} + x_{n-1}^{\text{ex}} = -(\delta_{n+1} - 2\delta_n + \delta_{n-1}) + \Delta_t^2 a_n + O(\Delta_t^4)$$

Here we see discrete versions of second derivatives

Recall: For any function f(t) defined on the time grid:

$$f_{n+1} = f_n + \Delta_t \dot{f}_n + \frac{1}{2} \Delta_t^2 \ddot{f}_n + \frac{1}{6} \Delta_t^3 \ddot{f}_n + O(\Delta_t^4)$$

$$f_{n-1} = f_n - \Delta_t \dot{f}_n + \frac{1}{2} \Delta_t^2 \ddot{f}_n - \frac{1}{6} \Delta_t^3 \ddot{f}_n + O(\Delta_t^4)$$

add these:

$$f_{n+1} - 2f_n + f_{n-1} = \Delta_t^2 \ddot{f}_n + O(\Delta_t^4)$$

$$(x_{n+1}^{\text{ex}} - 2x_n^{\text{ex}} + x_{n-1}^{\text{ex}})/\Delta_t^2 = -(\delta_{n+1} - 2\delta_n + \delta_{n-1})/\Delta_t^2 + a_n + O(\Delta_t^2)$$
$$(f_{n+1} - 2f_n + f_{n-1})/\Delta_t^2 = \ddot{f}_n + O(\Delta_t^2)$$

Replace discrete time derivatives by continuum versions:

$$\ddot{x}^{\text{ex}}(t) = -\ddot{\delta}(t) + a(t) + O(\Delta_t^2)$$

The exact solution satisfies

$$\ddot{x}^{\rm ex}(t) = a(t)$$

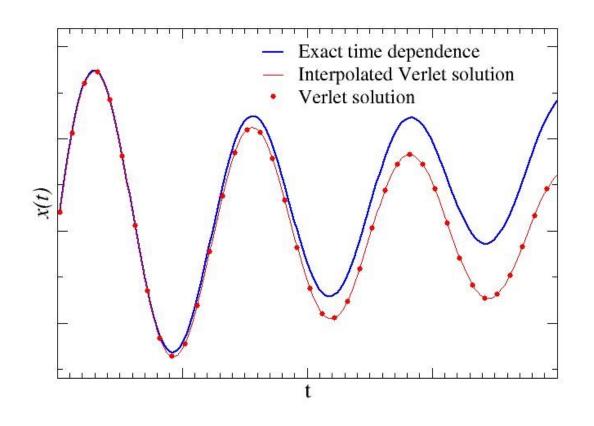
and we are left with

$$\ddot{\delta}(t) = O(\Delta_t^2)$$

Why is it OK to get rid of the discretization here?

We can always construct a smooth interpolation between the discrete values x_n

- e.g., a high-order polynomial



The equation for the error is rather incomplete

$$\ddot{\delta}(t) = O(\Delta_t^2)$$

We can also write it as an actual equation

$$\ddot{\delta}(t) = \Delta_t^2 g(t)$$

though we do not know anything about the function g(t)

For the accumulated error at time t=T we have to integrate

$$\delta(T) = \int_0^T \dot{\delta}(t)dt = \int_0^T \int_0^t \ddot{\delta}(t')dt' = \Delta_t^2 \int_0^T \int_0^t g(t')dt'$$

We cannot go any further without making assumptions for g(t)

- g(t) could oscillate around 0, leading to cancelations and small errors
 - this is the case for periodic motion
- g(t) could increase rapidly as t increases, causing large errors
 - the solution then likely goes completely bad after some time
- g(t) could be roughly time independent, leading to: $\delta(T) = O(\Delta_t^2 T^2)$
 - may be common for "well behaved" regular, non-periodic motion

Note: Δ_t must be sufficiently small for the above arguments to be valid

Leapfrog/Verlet method including damping

We assumed velocity-independent force (acceleration) in

$$v_{n+1/2}=v_{n-1/2}+\Delta_t a_n$$
 we do not have \emph{v}_n for \emph{a}_n = $\emph{a}(\emph{x}_n,\emph{v}_n,\emph{t})$
$$x_{n+1}=x_n+\Delta_t v_{n+1/2}$$

We can still use this form, with $a_n \rightarrow a(x_n, v_{n-1/2}, t)$, where error is $O(\Delta_t)$

- x error is then $O(\Delta_t^3)$ instead of $O(\Delta_t^4)$ [see by expanding a(v) in v]

To do better, first separate out dissipative part of force:

$$a(x, v, t) = \frac{1}{m} [F(x, t) - G(v)]$$

Consider the approximation

$$a(x_n, v_n, t_n) \approx [F(x_n, t_n) - G(v_{n-1/2})]/m$$

and use this for intermediate (^) velocity and position:

$$\hat{v}_{n+1/2} = v_{n-1/2} + \Delta_t [F(x_n, t_n) - G(v_{n-1/2})]/m$$
 $\hat{x}_{n+1} = x_n + \Delta_t \hat{v}_{n+1/2}$ has O(Δ_t 3) error

Then we can obtain v_n with $O(\Delta_t^2)$ error: $v_n = (\hat{x}_{n+1} - x_{n-1})/(2\Delta_t)$

$$v_n = (\hat{x}_{n+1} - x_{n-1})/(2\Delta_t) + O(\Delta_t^2)$$

Now we can use this in the acceleration $a_n(x_n,v_n,t)$; $O(\Delta_t^2)$ error Summary of procedure:

$$\begin{split} \hat{v}_{n+1/2} &= v_{n-1/2} + \Delta_t [F(x_n,t_n) - G(v_{n-1/2})]/m \\ \hat{x}_{n+1} &= x_n + \Delta_t \hat{v}_{n+1/2} \\ v_n &= (\hat{x}_{n+1} - x_{n-1})/(2\Delta_t) \\ v_{n+1/2} &= v_{n-1/2} + \Delta_t a_n \\ x_{n+1} &= x_n + \Delta_t v_{n+1/2} \end{split} \qquad \text{vn used here in an} \end{split}$$

More than twice as much work as the standard Leapfrog method

$$v_{n+1/2} = v_{n-1/2} + \Delta_t a_n$$
$$x_{n+1} = x_n + \Delta_t v_{n+1/2}$$

but the $O(\Delta_t^4)$ error is now maintained (work pays off)

Test by running friction.ipynb on the web site (tomorrow's discussion)

Runge-Kutta (RK) Method

A classic method with very small step error; $O(\Delta_t^5)$ Let's first apply it to a single 1st-order equation:

$$\dot{x}(t) = f[x(t), t]$$

It's instructive to first look at a simpler method with $O(\Delta_t^3)$ error 2nd-order RK method

Apply the mid-point rule (recall from numerical integration)

$$x_{n+1} = x_n + \int_{t_n}^{t_{n+1}} f[x(t), t] dt = \Delta_t f[x(t_{n+1/2}), t_{n+1/2}] + O(\Delta_t^3)$$

But here we do not have $x(t_{n+1/2}) = x_{n+1/2}$

We can approximate it using the first-order form

$$\hat{x}_{n+1/2} = x_n + \frac{\Delta_t}{2} f(x_n, t_n)$$
 ^ indicates an intermediate value of x. Error is $O(\Delta_t^2)$

- series expand f() in the integration formula to see that

$$x_{n+1} = x_n + \Delta_t f(\hat{x}_{n+1/2}, t_{n+1/2}) + O(\Delta_t^3)$$

Illustrates the use of intermediate values with larger error

4th-order RK method

Use Simpson's formula:

$$x_{n+1} = x_n + \int_{t_n}^{t_{n+1}} f[x(t), t] dt = x_n + \frac{\Delta_t}{6} (f_n + 4f_{n+1/2} + f_{n+1}) + O(\Delta_t^5)$$

where we need to find an approximation to $f_{n+1/2}$ and f_{n+1} with $O(\Delta_t^4)$ errors The way to do this is a bit involved/obscure...

$$\hat{x}_{n+1/2} = x_n + \Delta_t f(x_n, t_n)/2$$

first intermediate approximation

$$\hat{x}'_{n+1/2} = x_n + \Delta_t f(\hat{x}_{n+1/2}, t_{n+1/2})/2$$

improved intermediate approximation

similar for x_{n+1} . Scheme boils down to evaluating these:

$$k_1 = \Delta_t f(x_n, t_n)$$

$$k_2 = \Delta_t f(x_n + k_1/2, t_{n+1/2}),$$

$$k_3 = \Delta_t f(x_n + k_2/2, t_{n+1/2})$$

$$k_4 = \Delta_t f(x_n + k_3, t_{n+1})$$

and then

$$x_{n+1} = x_n + \frac{1}{6}(k1 + 2k_2 + 2k_3 + k_4)$$
 with $O(\Delta_t^5)$ step error

RK method for two coupled 1st-order equations

$$\dot{x}(t) = f(x, y, t) \quad \dot{y}(t) = g(x, y, t)$$

Simple generalization of the previous case:

$$egin{array}{lll} k_1&=&\Delta_t f(x_n,y_n,t_n),\ l_1&=&\Delta_t g(x_n,y_n,t_n),\ k_2&=&\Delta_t f(x_n+k_1/2,y_n+l_1/2,t_{n+1/2}),\ j_2&=&\Delta_t g(x_n+k_1/2,y_n+l_1/2,t_{n+1/2}),\ k_3&=&\Delta_t f(x_n+k_2/2,y_n+l_2/2,t_{n+1/2}),\ l_3&=&\Delta_t g(x_n+k_2/2,y_n+l_2/2,t_{n+1/2}),\ k_4&=&\Delta_t f(x_n+k_3,y_n+l_3,t_{n+1}),\ l_4&=&\Delta_t g(x_n+k_3,y_n+l_3,t_{n+1}),\ x_{n+1}&=&x_n+rac{1}{6}(k_1+2k_2+2k_3+k_4),\ y_{n+1}&=&y_n+rac{1}{6}(l_1+2l_2+2l_3+l_4), \end{array}$$

$k_1 = \Delta_t f(x_n, t_n)$ $k_2 = \Delta_t f(x_n + k_1/2, t_{n+1/2}),$ $k_3 = \Delta_t f(x_n + k_2/2, t_{n+1/2})$ $k_4 = \Delta_t f(x_n + k_3, t_{n+1})$ $x_{n+1} = x_n + \frac{1}{6}(k_1 + 2k_2 + 2k_3 + k_4)$

This is more general than Newton's equations

$$\dot{x} = y \ [y = v], \quad \dot{y} = a(x, y, t)$$

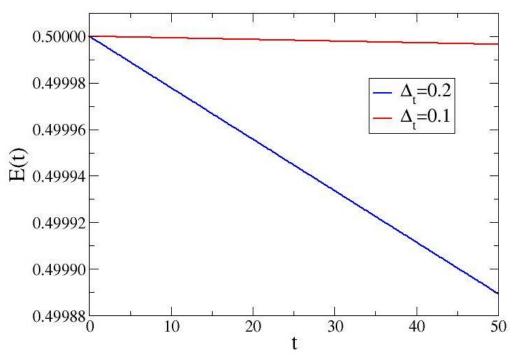
RK formulas for equations of motion

$$egin{array}{lll} k_1&=&\Delta_t a(x_n,v_n,t_n),\ l_1&=&\Delta_t v_n,\ k_2&=&\Delta_t a(x_n+l_1/2,v_n+k_1/2,t_{n+1/2}),\ l_2&=&\Delta_t (v_n+k_1/2),\ k_3&=&\Delta_t a(x_n+l_2/2,v_n+k_2/2,t_{n+1/2}),\ l_3&=&\Delta_t (v_n+k_2/2),\ k_4&=&\Delta_t a(x_n+l_3,v_n+k_3,t_{n+1}),\ l_4&=&\Delta_t (v_n+k_3),\ v_{n+1}&=&y_n+rac{1}{6}(k_1+2k_2+2k_3+k_4),\ x_{n+1}&=&x_n+rac{1}{6}(l_1+2l_2+2l_3+l_4). \end{array}$$

Friction (dissipative, v-dependent forces) can be included directly here

Test of the 4th-order RK method

Same harmonic oscillator as before

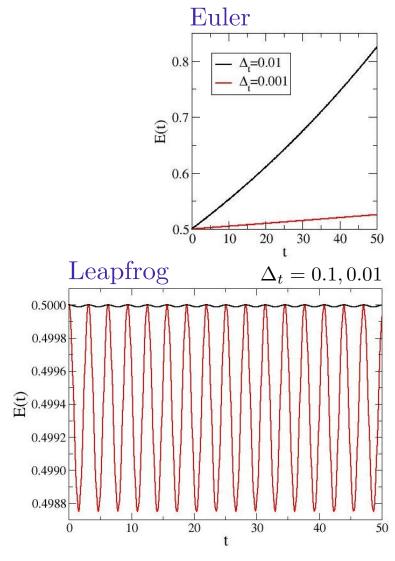


The energy error is not bounded

- not even for periodic motion

No time-reversal symmetry

- can be important in some applications



May be better than Leapfrog when E is not conserved (damping, driving)