# Report for Worksheet 1: Integrators

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## **Contents**

1	Can	nonball
	1.1	Simulating a cannonball
	1.2	Influence of friction and wind
2	Sola	ar system
	2.1	
	2.2	Integrators
		2.2.1 Velocity Verlet algorithm
		2.2.2 Verlet algorithm
		2.2.3 Implementation of the symplectic Euler algorithm
		2.2.4 Implementation of the Velocity Verlet algorithm
	2.3	Long-term stability

## 1 Cannonball

## 1.1 Simulating a cannonball

The function for the (constant) gravitational force can be implemented in the following way in Python, it returns the force as a vector:

```
def force(mass, gravity):
    return np.array([0.0, -mass * gravity])
```

The Euler scheme is implemented in this fashion:

```
def step_euler(x, v, dt, mass, gravity, f):
    x += v * dt
    v += f / mass * dt
    return x, v
```

It is crucial to first update the positions  $\mathbf{x}$  and then update the velocites  $\mathbf{v}$ . To simulate the cannonball until it hits the ground, a while-loop is used.

A plot of the simulated trajectory y(x) is shown in Figure 1. As we would expect from the analytical solution

$$y(x) = x - \frac{gx^2}{2v_0^2},\tag{1}$$

the trajectory looks like a parabola.

#### 1.2 Influence of friction and wind

To account for friction, the function which calculates the force is modified in the following way:

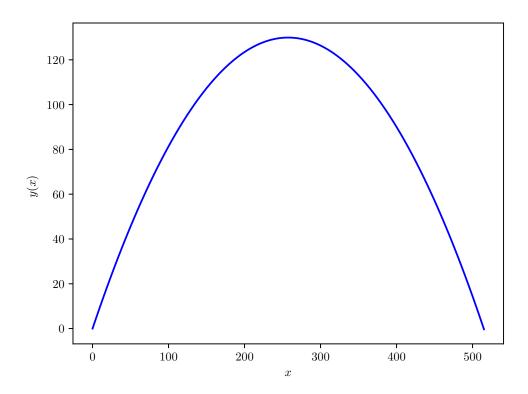


Figure 1: Simulated trajectory y(x) for the system without friction. The used integrator is the Euler scheme.

```
def force(mass, gravity, v, gamma, v_0):
    ret = np.array([0.0, -mass * gravity])
    ret -= gamma * (v - v_0)
    return ret
```

Because the force is now not constant anymore along the trajectory (it varies with the velocity  $\mathbf{v}$ ), the function for the Euler step has to be modified as well:

```
def step(x, v, dt, mass, gravity, gamma, v_0):
    x += v * dt
    f = force(mass, gravity, v, gamma, v_0)
    v += f / mass * dt
    return x, v
```

In contrast to the previous task, the force is now evaluated every time the Euler step is called.

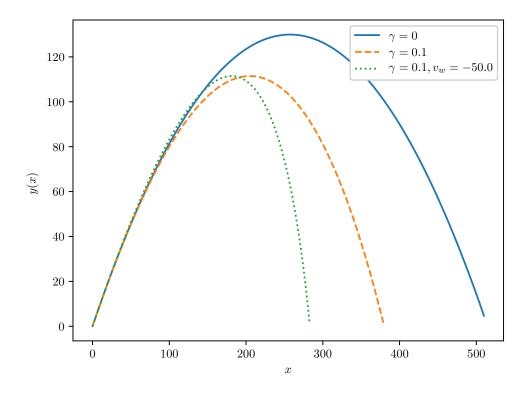


Figure 2: Simulated trajectories y(x) for a friction coefficient  $\gamma = 0.1$  and different values of the wind speed  $v_w$ . The used integrator is the Euler scheme.

Figure 2 shows simulated trajectories for the system without friction as well as for a system with friction coefficient  $\gamma=0.1$  and wind speeds of  $v_w=0,-50.0$ . Comparing the trajectory of the cannonball without friction to the ones with friction, we can easily see that the friction leads to both a decreased maximum height  $y_{\rm max}$  and a decreased range  $x_{\rm max}$ . This is caused by the disspation of energy through the non-conservative friction force. We can also identify that the negative wind speed  $v_w=-50.0$  leads to an even larger decrease of the range  $x_{\rm max}$ , this is also expected, because the friction force is proportional to the relative velocity of the air and the cannonball. Because the wind blows in the x-direction only, the maximum height  $y_{\rm max}$  is the same for both cases.

In Figure 3 we see multiple trajectories for  $\gamma=0.1$  and different wind speeds. As the wind speed becomes more negative, the range of the cannonball becomes smaller because the friction increases. For  $v_w\approx 200$  it hits the ground at its starting point.

# 2 Solar system

#### 2.1 Simulating the solar system with the Euler scheme

The gravitational force between two particles is calculated using this function:

```
def force(r_ij, m_i, m_j, g):
    return - g * m_i * m_j * r_ij / np.linalg.norm(r_ij) ** 3
```

To calculate all the forces on all the particles, the following function is used.

```
def forces(x, masses, g):
    ret = np.zeros(x.shape)

for i in range(len(x)):
    for j in range(i + 1, len(x)):
```

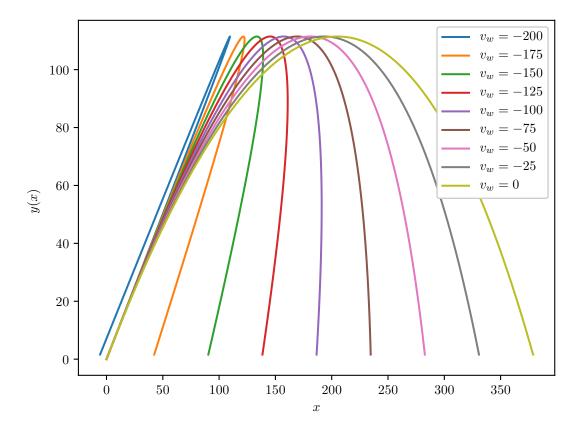


Figure 3: Simulated trajectories y(x) for different values of the friction coefficient  $\gamma$  and the wind speed  $v_w$ . For  $v_w = -200$ , the cannonball hits the ground closely to the starting point. The used integrator is the Euler scheme.

```
f = force(x[j] - x[i], masses[i], masses[j], g)
    ret[i] -= f
    ret[j] += f

return ret
```

The function returns an array which contains all forces.

A small modification of the Euler step is necessary because the particles have different masses:

```
def step_euler(x, v, dt, mass, g):
    x += v * dt
    f = forces(x, mass, g)
    # calculate acceleration per coordinate dimension
    f[:, 0] /= mass
    f[:, 1] /= mass
    v += f * dt
    return x, v
```

?? shows the numerically calculated trajectories of the celestial bodies over a time of one year for a time step of  $\Delta t = 10^{-4}$ . We can see that the earth performs almost one orbit around the sun. Figure 5 shows the trajectory of the moon in the rest frame of the earth for different time steps. The trajectories are not closed, for the smaller time step the orbits are closer together, but still not identical.

In simulations with many particles, the computationally most expensive step is the evaluation of the forces  $\mathbf{F}_{ij}$ . For a system of n particles, the complexity of this task scales like  $\mathcal{O}(n^2)$  because

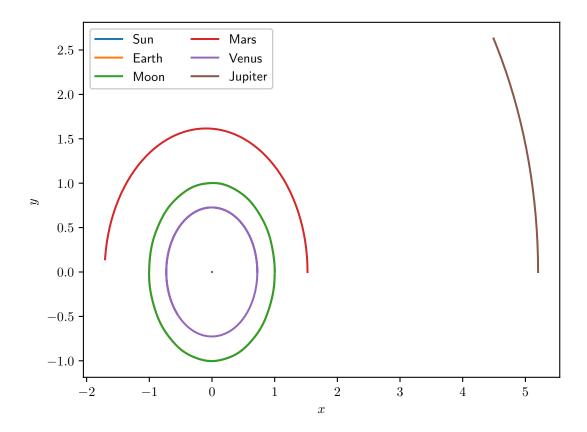


Figure 4: Trajectories of the celestial bodies over one year.

the pairwise forces  $\mathbf{F}_{ij}$  have to be evaluated for every of the n(n-1) possible combination of i, j (or at least half of them using Newton's third law).

### 2.2 Integrators

#### 2.2.1 Velocity Verlet algorithm

The Velocity Verlet algorithm can be derived in the following way: For the positions  $\mathbf{x}$ , we perform a Taylor expansion up to the second order in  $\Delta t$ :

$$\mathbf{x}(t + \Delta t) = \mathbf{x}(t) + \dot{\mathbf{x}}(t) \cdot \Delta t + \frac{\ddot{\mathbf{x}}(t)}{2} \cdot \Delta t^2 + \mathcal{O}(\Delta t^3)$$

$$= \mathbf{x}(t) + \mathbf{v}(t) \cdot \Delta t + \frac{\mathbf{a}(t)}{2} \cdot \Delta t^2 + \mathcal{O}(\Delta t^3)$$
(2)

For the velocites  $\mathbf{v}$ , we also perform a Taylor expansion up to the second order:

$$\mathbf{v}(t + \Delta t) = \mathbf{v}(t) + \dot{\mathbf{v}}(t) \cdot \Delta t + \frac{\ddot{\mathbf{v}}(t)}{2} \cdot \Delta t^2 + \mathcal{O}(\Delta t^3)$$

$$= \mathbf{v}(t) + \mathbf{a}(t) \cdot \Delta t + \frac{\ddot{\mathbf{v}}(t)}{2} \cdot \Delta t^2 + \mathcal{O}(\Delta t^3)$$
(3)

To get an expression for  $\ddot{\mathbf{v}}(t)$ , we perform a Taylor expansion of  $\dot{\mathbf{v}}(t+\Delta t)$ :

$$\dot{\mathbf{v}}(t + \Delta t) = \dot{\mathbf{v}}(t) + \ddot{\mathbf{v}}(t) \cdot \Delta t + \mathcal{O}(\Delta t^2) \tag{4}$$

We can solve this expression for  $\ddot{\mathbf{v}}(t)$ :

$$\ddot{\mathbf{v}}(t) = \frac{\dot{\mathbf{v}}(t + \Delta t) - \dot{\mathbf{v}}(t)}{\Delta t} + \mathcal{O}(\Delta t) = \frac{\mathbf{a}(t + \Delta t) - \mathbf{a}(t)}{\Delta t} + \mathcal{O}(\Delta t)$$
 (5)

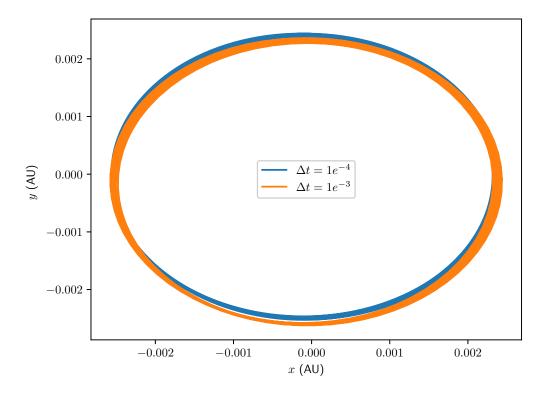


Figure 5: Trajectory of the moon in the rest frame of the earth for different time steps.

Plugging  $\ddot{\mathbf{v}}(t)$  into Equation 3, we get

$$\mathbf{v}(t + \Delta t) = \mathbf{v}(t) + \frac{\mathbf{a}(t + \Delta t) + \mathbf{a}(t)}{2} \cdot \Delta t^2 + \mathcal{O}(\Delta t^3)$$
(6)

Equation 2 and Equation 6 define the Velocity Verlet algorithm.

#### 2.2.2 Verlet algorithm

The Verlet algorithm which was presented in the lecture can also be derived from the Velocity Verlet algorithm. Using Equation 2, we can write  $\mathbf{x}(t+2\Delta t)$  as

$$\mathbf{x}(t+2\Delta t) = \mathbf{x}(t+\Delta t) + \mathbf{v}(t+\Delta t) \cdot \Delta t + \frac{\mathbf{a}(t+\Delta t)}{2} \cdot \Delta t^2 + \mathcal{O}(\Delta t^3)$$
 (7)

and  $\mathbf{x}(t)$  as

$$\mathbf{x}(t) = \mathbf{x}(t + \Delta t) - \mathbf{v}(t) \cdot \Delta t - \frac{\mathbf{a}(t)}{2} \cdot \Delta t^2 + \mathcal{O}(\Delta t^3). \tag{8}$$

Adding these equations results in

$$\mathbf{x}(t+2\Delta t) + \mathbf{x}(t) = \mathbf{x}(t+\Delta t) + \mathbf{v}(t+\Delta t) \cdot \Delta t + \frac{\mathbf{a}(t+\Delta t)}{2} \cdot \Delta t^{2}$$

$$+ \mathbf{x}(t+\Delta t) - \mathbf{v}(t) \cdot \Delta t - \frac{\mathbf{a}(t)}{2} \cdot \Delta t^{2} + \mathcal{O}(\Delta t^{3})$$

$$= 2\mathbf{x}(t+\Delta t) + \mathbf{a}(t+\Delta t) \cdot \Delta t^{2} + \mathcal{O}(\Delta t^{3}).$$
(10)

where we used the relation

$$\mathbf{v}(t + \Delta t) - \mathbf{v}(t) = \frac{\mathbf{a}(t) + \mathbf{a}(t + \Delta t)}{2} \cdot \Delta t$$
(11)

which can be derived from Equation 6. Isolating  $\mathbf{x}(t + \Delta t)$  and shifting t by  $-\Delta t$  results in the Verlet algorithm

$$\mathbf{x}(t + \Delta t) = 2\mathbf{x}(t) - \mathbf{x}(t - \Delta t) + \mathbf{a}(t + \Delta t) \cdot \Delta t^{2}$$
(12)

As we can see in Equation 12, the Verlet algorithm needs the positions  $\mathbf{x}(t)$  and  $\mathbf{x}(t - \Delta t)$  to calculate the position  $\mathbf{x}(t + \Delta t)$ . However, the initial conditions only include the position at exactly one point in time  $t_0$ , this means that the initial conditions are not sufficient to solve the problem with the Verlet algorithm.

#### 2.2.3 Implementation of the symplectic Euler algorithm

The symplectic Euler scheme can be implemented in the following way:

```
def step_euler_symplectic(x, v, dt, mass, g):
    f = forces(x, mass, g)
    f[:, 0] /= mass
    f[:, 1] /= mass
    v += f * dt
    x += v * dt
    return x, v
```

In contrast to the Euler scheme used above, velocites are updated first.

#### 2.2.4 Implementation of the Velocity Verlet algorithm

The Velocity Verlet algorithm is implemented in this fashion:

```
def step_velocity_verlet(x, v, dt, mass, g):
    f = forces(x, mass, g)
    f[:, 0] /= mass
    f[:, 1] /= mass # now it's acceleration

x += v * dt + f / 2 * dt * dt

# a( t + dt )
    f_t = forces(x, mass, g)
    f_t[:, 0] /= mass
    f_t[:, 1] /= mass # now it's acceleration

v += (f + f_t) / 2 * dt

return x, v
```

Figure 6 shows the trajectory of the moon for different integrators with a large time step of  $\Delta t = 0.01$ . We can see that the trajectory for Euler scheme is unstable and the distance grows rapidly. The other two integrators seem to do fine for this simulation time. The trajectory obtained with the velocity Verlet algorithm is closer together which indicates that the integration error is smaller.

## 2.3 Long-term stability

The distance between earth and moon diverges after a certain number of timesteps due to the error introduced with only a hundered force recalculations per year. This leads to breaking the gravitational bonds between them after a certain number of steps which depend on the order of the integration algorithm. This is demonstrated by the symplectic Euler diverging much later than the standard Euler and the Velocity Verlet not showing divergence after a ten year simulation period of thousand steps in total, see Figure 7.

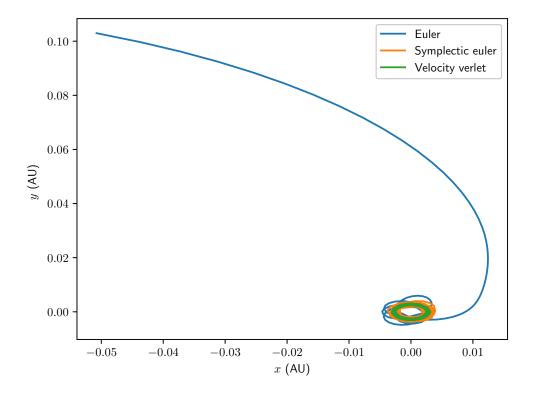


Figure 6: Trajectory of the moon over one year in the rest frame of the earth for different integrators.

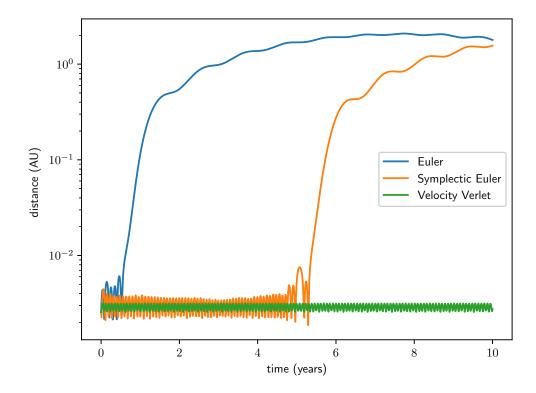


Figure 7: Distance between earth and moon over a simulation time of 10 years with 100 steps per year using different integrators.