#### ex01

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### Numerical Methods in Plasma Astrophysics

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Exercise 1: N-Body Rubber Ducks

```
[121]: import time
       import numpy as np
       import matplotlib.pyplot as plt
       from matplotlib.animation import FuncAnimation
       %matplotlib inline
       np.random.seed(0)
       def Calculate_Temperature(vel,N):
           real_vel = vel.copy()
           v_squared = np.sum(real_vel*real_vel, axis=1)
           v_mean = np.mean(v_squared)
           ene_kin_aver = 0.5*v_mean
           temperature = ene_kin_aver/3
           return ene_kin_aver,temperature
       def compute_force(x):
           rm = 2
           n = x.shape[0]
           a = np.zeros(x.shape)
           energy_pot = np.zeros(x.shape)
           for i in range(n):
               for j in range(n):
                   if i==j:
                       continue
                   r_{vec} = x[i] - x[j]
                   for 1 in range(2): # Periodic interactions
                       if (np.abs(r_vec[1])>10):
                           r_{vec}[1] = r_{vec}[1] - np.copysign(20.0, r_{vec}[1])
                   r = np.sqrt(r_vec@r_vec)
                   if (r<10):
```

```
phi = (rm/r)**12 - 2*(rm/r)**6
                dphi = 12*rm**12/r**14 - 12*rm**6/r**8
                a[i] += dphi*r_vec
                a[j] -= dphi*r_vec
                energy_pot[i] += 0.5*phi
                energy_pot[j] += 0.5*phi
    return a, np.mean(energy_pot)
def solve_nbody(x, v, mu, dt, steps, method):
    n = x.shape[0]
    traj = np.zeros((steps, n, 2))
    temps = np.zeros(steps)
    energy_kin = np.zeros(steps)
    energy_pot = np.zeros(steps)
    for s in range(steps):
        for i in range(2):
            period = np.where(x[:,i] > 10)
            x[period,i]=x[period,i]-20
            period = np.where(x[:,i] < -10)
            x[period,i]=x[period,i]+20
        if method=="Leapfrog":
            x half = x + v*dt/2
            a, energy_pot[s] = compute_force(x_half)
            energy_kin[s],temps[s] = Calculate_Temperature(v, n)
            v = v + (a - mu*v)*dt
            x = x half + dt/2*v
        if method=="Euler":
            x = x + v*dt
            a, energy_pot[s] = compute_force(x)
            energy_kin[s],temps[s] = Calculate_Temperature(v, n)
            v = v + a*dt
        traj[s] = x
    return traj, energy_kin, energy_pot, temps
```

# a) Development of rubber duck population

The temperature is dependend on the standard deviation of the velocity distribution, so starting with different standard deviation is equivalent to starting with different temperatures

```
[91]: n = 5**2

mu = 0.0

dt = 0.01

steps = 1000

boxsize = 10
```

```
lin = np.linspace(-9,9,int(np.sqrt(n)))
X,Y = np.meshgrid(lin,lin)
x = np.vstack([X.ravel(), Y.ravel()]).T
```

```
[93]: v = np.random.normal(0,5.0,(n,2))

traj, energy_kin, energy_pot, temps = solve_nbody(x, v, mu, dt, steps, u

→"Leapfrog")

plt.plot(x[:,0], x[:,1], "o", label="Initial positions")

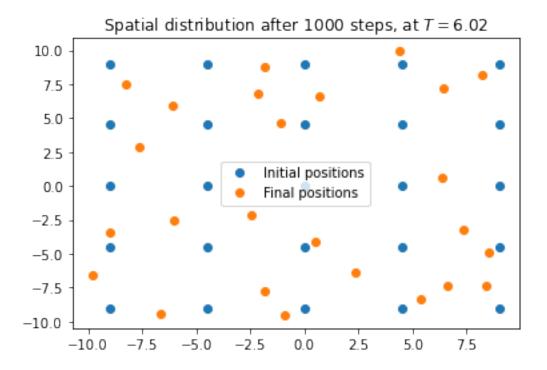
plt.plot(traj[-1,:,0], traj[-1,:,1], "o", label="Final positions")

plt.title("Spatial distribution after {} steps, at $T=${:.2f}}".format(steps, u

→temps[-1]))

plt.legend()
```

[93]: <matplotlib.legend.Legend at 0x7fc4a6798190>



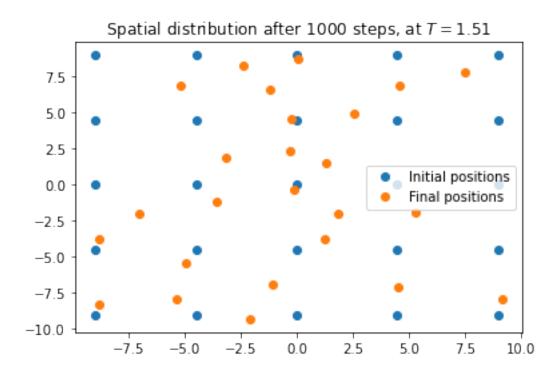
```
[94]: v = np.random.normal(0,2.0,(n,2))
traj, energy_kin, energy_pot, temps = solve_nbody(x, v, mu, dt, steps, u

→"Leapfrog")
plt.plot(x[:,0], x[:,1], "o", label="Initial positions")
plt.plot(traj[-1,:,0], traj[-1,:,1], "o", label="Final positions")
plt.title("Spatial distribution after {} steps, at $T=${:.2f}".format(steps, u

→temps[-1]))
```

```
plt.legend()
```

#### [94]: <matplotlib.legend.Legend at 0x7fc4a6af5160>



```
[95]: v = np.random.normal(0,1.0,(n,2))

traj, energy_kin, energy_pot, temps = solve_nbody(x, v, mu, dt, steps,

→"Leapfrog")

plt.plot(x[:,0], x[:,1], "o", label="Initial positions")

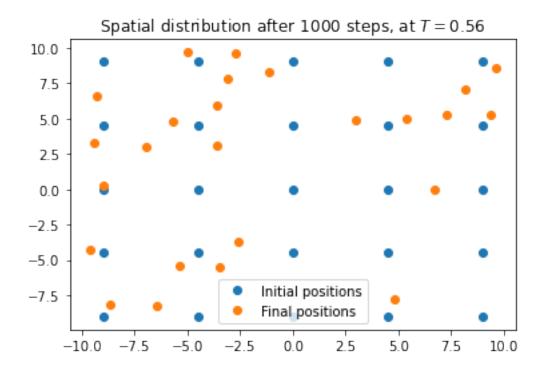
plt.plot(traj[-1,:,0], traj[-1,:,1],"o", label="Final positions")

plt.title("Spatial distribution after {} steps, at $T=${:.2f}".format(steps,

→temps[-1]))

plt.legend()
```

[95]: <matplotlib.legend.Legend at 0x7fc4a6c28670>

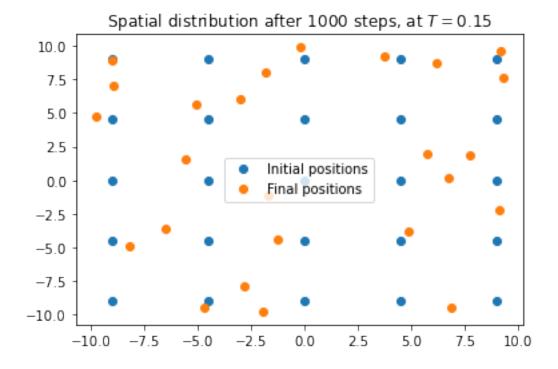


```
[96]: v = np.random.normal(0,0.1,(n,2))
traj, energy_kin, energy_pot, temps = solve_nbody(x, v, mu, dt, steps,

→"Leapfrog")
plt.plot(x[:,0], x[:,1], "o", label="Initial positions")
plt.plot(traj[-1,:,0], traj[-1,:,1],"o", label="Final positions")
plt.title("Spatial distribution after {} steps, at $T=${:.2f}".format(steps,

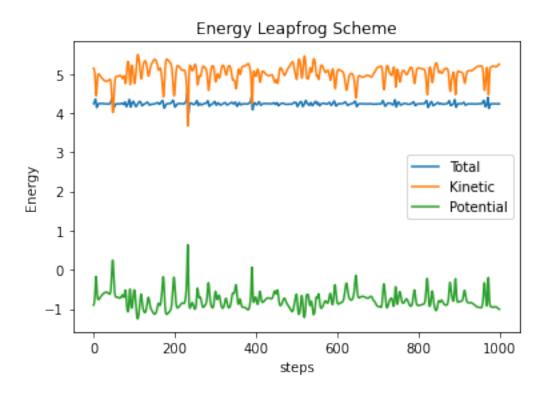
→temps[-1]))
plt.legend()
```

[96]: <matplotlib.legend.Legend at 0x7fc4a6856df0>

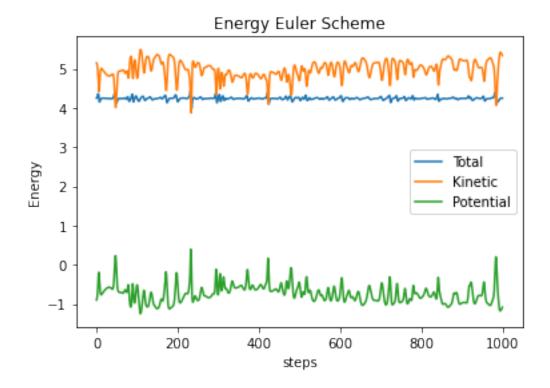


For high temperatures, no pattern in the arrangement of the particles can be observed. However, with decreasing temperature we can see a progressive formation of clusters of particles which "stick" together. At lowest temperature the particles are making chains and crystal-like structures.

### b) Total energy developement

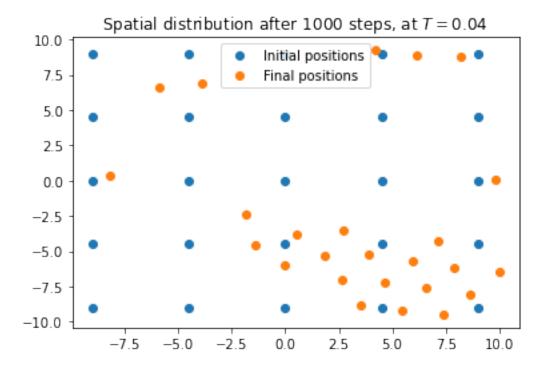


```
[87]: traj, energy_kin, energy_pot, temps = solve_nbody(x, v, mu, dt, steps, "Euler")
    plt.plot((energy_kin+energy_pot), label="Total")
    plt.plot(energy_kin, label="Kinetic")
    plt.plot(energy_pot, label="Potential")
    plt.title("Energy Euler Scheme")
    plt.xlabel("steps")
    plt.ylabel("Energy")
    plt.legend()
    plt.show()
```

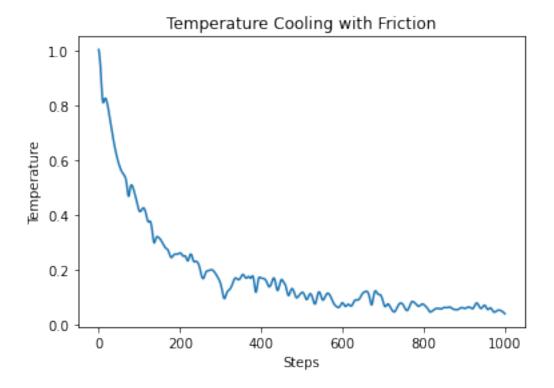


The total energy, the sum of kinetic and potential energy, stays more or less constant over time as expected for a energy conserving system. The Leapfrog scheme seems to perform slightly better than the Euler schmeme, which tends to have bigger oscilations in the energies.

#### c) Cooling



```
[101]: plt.title("Temperature Cooling with Friction")
   plt.plot(temps)
   plt.xlabel("Steps")
   plt.ylabel("Temperature")
   plt.show()
```



As the temperature falls, the ducks start to make crystal like patters (hexagonal in our case, because of optimal distribution in a plane). This is because the lennard-jones potential has a minimum, the preferred distance  $r_m$ .

## Additional: Animation of the crystal formation when cooling down the system

```
[17]: fig = plt.figure()
    ax = plt.axes(xlim=(-10, 10), ylim=(-10, 10))

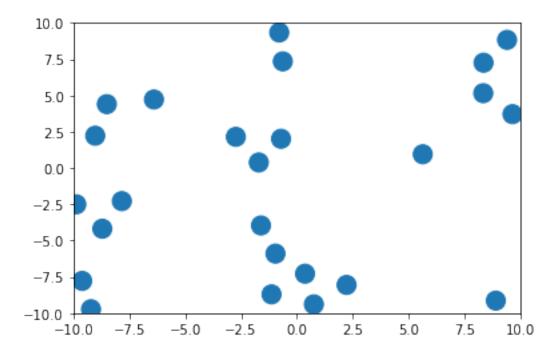
n = 10
v = np.random.normal(0,1,(n,2))
x = np.random.uniform(0,10,(n,2))

line = ax.scatter(traj[0,:,0], traj[0,:,1], s=np.ones(traj[0,:,0].shape)*200)

dt = 0.01
def init():
    line.set_offsets(np.c_[x[:,0], x[:,1]])
    return line,
def animate(i):
    x = traj[i]
    line.set_offsets(np.c_[x[:,0], x[:,1]])
```

```
return line,
anim = FuncAnimation(fig, animate, frames=1000, interval=20)
anim.save('lennard_jones_cooling.gif',writer='imagemagick')
```

MovieWriter imagemagick unavailable; using Pillow instead.



# 1 d) Plot of Runtime for Euler and Leapfrog

```
[109]: def eval_time(n):
    #n = 5**2
    mu = 0.0
    dt = 0.01
    steps = 10
    boxsize = 10

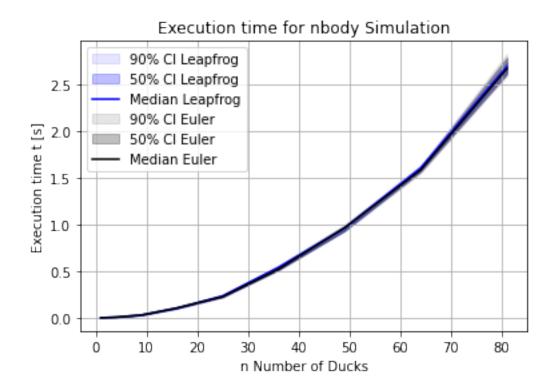
lin = np.linspace(-9,9,int(np.sqrt(n)))
    X,Y = np.meshgrid(lin,lin)

x = np.vstack([X.ravel(), Y.ravel()]).T
    v = np.random.normal(0,0.1,(n,2))

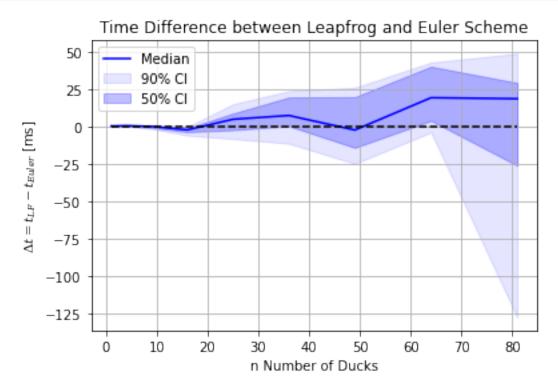
start_time = time.time()
```

```
traj, energy_kin, energy_pot, temps = solve_nbody(x, v, mu, dt, steps, u
       end_time = time.time()
          time_lf = end_time - start_time
          start time = time.time()
          traj, energy_kin, energy_pot, temps = solve_nbody(x, v, mu, dt, steps,_
       end_time = time.time()
          time_euler = end_time - start_time
          return np.array([time_lf, time_euler])
[113]: ns = [i**2 for i in range(1,10)]
      times = []
      repeat = 10
      print(repeat)
      for i in range(repeat):
          print("Current round:", i)
          times.append(np.array([eval_time(n) for n in ns]))
      times = np.array(times)
      times lf = times[:,:,0]
      times_euler = times[:,:,1]
      10
      Current round: 0
      Current round: 1
      Current round: 2
      Current round: 3
      Current round: 4
      Current round: 5
      Current round: 6
      Current round: 7
      Current round: 8
      Current round: 9
[122]: plt.fill_between(ns, np.percentile(times_lf, axis=0, q=[10]).ravel()[:],
                           np.percentile(times_lf, axis=0, q=[90]).ravel()[:],
                            alpha = 0.1, color='b', label='90% CI Leapfrog');
      plt.fill_between(ns, np.percentile(times_lf, axis=0, q=[25]).ravel()[:],
                           np.percentile(times_lf, axis=0, q=[75]).ravel()[:],
                           alpha = 0.25, color='b', label='50% CI Leapfrog');
      plt.plot(ns, np.mean(times_lf, axis=0)[:], color="blue", label='Medianu
       →Leapfrog')
```

[122]: Text(0.5, 1.0, 'Execution time for nbody Simulation')



```
[123]: delta_time = (times_lf-times_euler)*1e3
    plt.rcParams["mathtext.fontset"] = "dejavuserif"
    plt.plot(ns, np.median(delta_time, axis=0), color="blue", label="Median")
```



Leapfrog is a little big slower than the Euler scheme. The difference is however negligible compared to the overall runtimes of the algorithms.