```
diff --git a/cal_stepl_solution.py b/cal_stepl_solution.py
index 17d9dc2..e511adf 100644
--- a/ca1_step1_solution.py
+++ b/ca1_step1_solution.py
@@ -1,131 +1,151 @@
+# Please put everything in the docstring also the comments
 #88
         CA 1: STEP 1
 ''' Group 26:
    Anton Gusarov
    Martin Gustavsson
 import re
 import numpy as np
 import h5py
 import matplotlib
 from matplotlib import pyplot as plt
 from itertools import islice
 #%% Simply reading the first 2 lines:
 with open('cal_step1_input_data.txt', 'r', errors='strict') as file: # opener is a custom callable
     lines = list()
     for i in range(2):
         lines = file.readline()
 # %% Parse the main header of the file (first 2 lines):
 with open('cal_stepl_input_data.txt', 'r', errors='strict') as file:
     # read the header only i.e. the first 2 lines
     # reg. expr. used where assumed that ',' or ';' are the separators:
     lines = [re.split(r'[,;]', line)] for line in islice(file.readlines(), 2)]
 header_keys = tuple(i.strip() for i in lines[0]) # remove trailing spaces by .strip()
 header_keys = tuple(re.sub(r'[\s#)]','', i) for i in header_keys)
header_keys = tuple(re.sub(r'[(]','_', i) for i in header_keys)
 # %% remove all space characters from header line 2 and tx to float:
 header_values = [float(i.strip()) for i in lines[1]]
 header_values[0] = int(header_values[0])  # time_steps to be int
 header_values[4] = int(header_values[4]) # N_particles to be int
 # %% Create a dict {header_key -> header_value}:
header = {}
+# Please use pythonic looping with zip, i.e.
+# for key, value in zip(header_keys, header_values):
 for i in range(len(header keys)):
     header[header_keys[i]] = header_values[i] # build dict incrementally
 locals().update(header) # unzip the dict to a set of variables
 # %% You need to identify the start of each time step by looking for the
 # time-step header # time_step. And then parse the N_particles rows
 # of position and velocity data in each time-step block
 # First task: print headers for time_step:
 with open('cal_stepl_input_data.txt', 'r', errors='strict') as file:
     for line in islice(file.readlines(), 4):
         if '# time_step ' in line: print(line)
 # %%
 # Parse the data in each time-step block.
 # Store the data in a (three level) nested list called data.
 \# 1. The first level should be the time step index,
 # 2. the second level the particle index,
 \# 3. third level a list of the four values of the x, y position-
      and v_x, v_y velocitiy-components stored as floating point numbers (float).
 inner_list_particle = []
 data = []
 particle_count = 0
 # The parser:
 with open('cal_stepl_input_data.txt', 'r', errors='strict') as file:
     for line in islice(file.readlines(), 3, None, None): # start from trird line - skip the header
         if (not line) or line.isspace() : # skip empty line: no symbols or only spaces
             #print('empty string')
             continue
         if '# time_step ' in line:
             time\_step\_index = int(''.join(re.findall(r'\d',line))) # parsed time-step index
             #print('time step:', time_step_index, '\n')
         if re.search(r' \# ', line) is None: \# find a line without ' \# ' symbol i.e. with position and speeds numbers
             inner_list_particle.append([float(i.strip()) for i in re.split(r'[,;]', line)])
             particle_count += 1
```

```
# Please do not hard code the number of particles
             if particle_count == 400:
                  # This works, but is extra work for the computer
                  # Please do not copy the data and then remove the original
                  # Just pass the data along by appending it
                  # and set inner_list_particle to a new empty list []
                 data.append(inner_list_particle.copy())
                 inner list particle.clear()
                 particle_count = 0
print('data list size:', len(data), len(data[0]), len(data[0][0])) # check data list dimensions
 \# %% Convert a nested list to a multidimensional (NumPy) array
 data = np.array(data)
 print(type(data))
 print(data.dtype)
 print (data.shape)
 # %% Split position vectors and velocity vectors
 # To make the numerical analysis easier you want to store the final result in two separate ndarrays,
 \ensuremath{\text{\#}} one for the positions called R and one for the velocities called V
 R = data[:,:,:2]
 V = data[:,:,2:]
 print(R.shape)
print(V.shape)
 # %% To swap the order of the axes use the NumPy "transpose-like" manipulation routines:
 R = np.swapaxes(R, 1, 2)
 V = np.swapaxes(V, 1, 2)
 print(R.shape)
 print (V.shape)
 \# %% Plot the first time step with Matplotlib
 fig, axs = plt.subplots()
 axs.set_aspect(aspect=1)
axs.scatter(R[0, 0, :], R[0, 1, :], c='red', s=50.0, alpha=0.4, edgecolors='none')
 plt.xlim(-1, 1)
 plt.ylim(-1, 1)
 plt.show(fig)
 # %% Storing data using the hdf5 file format
 # use its h5py.File class together with the 'with'-statement to open the file ca1_step1_output_data.h5 file.
 # Use '.create_dataset' method to store the ndarrays R and V,
 \# and store the other parameters as attributes using the .attrs.create method.
 with h5py.File('cal_stepl_output_data.h5', 'w') as hfile:
     dset1 = hfile.create_dataset('R', data=R)
     dset1 = hfile.create_dataset('V', data=V)
     # Please store the attributes in the root of the hdf5 file not in dset1
     # using the function pointed out in the description
     dset1.attrs['N_particles'] = N_particles
     dset1.attrs['time_steps'] = time_steps
dset1.attrs['time_step_s'] = time_step_s
     dset1.attrs['radius'] = radius
     dset1.attrs['v_variance'] = v_variance
diff --git a/ca1_step3_solution.py b/ca1_step3_solution.py
index a8bfdff..502ecaa 100644
--- a/ca1_step3_solution.py
+++ b/ca1_step3_solution.py
@@ -1,216 +1,235 @@
+# use docstring here not comments
 #88
        CA 1: STEP 3
 import os
 import itertools
 import h5py
 from scipy import stats
 from scipy.spatial import distance
 import numpy as np
 {\tt from\ numpy\ import\ random}
 from matplotlib import pyplot as plt
 import matplotlib.animation as animation
 from scipy.spatial import distance
 from scipy.spatial import cKDTree
```

```
from ca1_step_2_solution import animate_particles
from cal_step_2_solution import plot_statistical_analysis
+# Please move all functions here and put the other code after them
# 1. Particle system initialization
N_particles = 400
sigma_v = 0.1
r0 = 2 * random.rand(2, N_particles) - 1
v0 = sigma_v * random.randn(2, N_particles)
print(r0.shape, v0.shape) # shape check
# sanity check plot of initial positions:
x0, y0 = r0[0, :], r0[1, :]
fig, axs = plt.subplots()
axs.set_aspect(aspect=1)
axs.scatter(x0, y0, color='green', s=50.0, alpha=0.4, edgecolors='none')
plt.xlim(-1.1, 1.1)
plt.ylim(-1.1, 1.1)
plt.show(fig)
# %%
# 2. Time propagation
time_step = 0.02 # 50 steps/sec
def time_propagation(r, v, time_step):
    r = r + v + time_step # broadcasting was used to implement this in a single line
    return r, v
r, v = time_propagation(r0, v0, time_step) # call the function
# sanity check plot of time propagation:
x, y = r[0, :], r[1, :]
fig, axs = plt.subplots()
axs.set_aspect(aspect=1)
axs.scatter(x, y, color='blue', s=50.0, alpha=0.4, edgecolors='none')
plt.xlim(-1.1, 1.1)
plt.ylim(-1.1, 1.1)
plt.show(fig)
# 3. Simulation function
# The function uses the time propagation function to compute
# the time evolution of the particles for a given number of 'time_steps'
\mbox{\#} and returns the positions R and velocities V of all particles for all time steps.
def simulate(r0, v0, time_step, time_steps, update_function, **kwargs):
     # Allocate R and V:
    R = np.empty((time_steps, r0.shape[0], r0.shape[1]))
    V = np.empty((time_steps, v0.shape[0], v0.shape[1]))
     \# Initialize R[0] and V[0] using r0 and v0:
    R[0,:,:] = r0
V[0,:,:] = v0
     # Loop over all time steps and call:
     for t in range(1, time_steps):
        R[t,:,:], V[t,:,:] = update_function(R[t-1,:,:], V[t-1,:,:], time_step, **kwargs)
    return R, V
simulation1 = simulate(
    r0, v0, time_step, time_steps=400,
    update_function=time_propagation)
# call the animation:
animate_particles(simulation1[0], time_step, filename='ca1_step3_movie1.mp4')
# 4. Hard wall boundary collisions
\# -- check if the position coordinates (x and y) are out of bounds and
 # -- the particle at the same time has a velocity component going in the wrong direction (relative to the "wall").
# -- In this case change the sign of the corresponding velocity component (v_x or v_y)
# Use numpy boolean arrays for indexing so that each line of code treats all particles at the same time.
\# Modifiy the velocity vector v in-place so that the function does not have to return anything.
def boundary_collisions(r, v):
     ^{\prime\prime\prime} Modifies the velocity vector according to the hard wall boundary conditions.
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# Please add the check of the velocity components
       # Without it particles tend to get stuck at the boundaries and even go outside it
      boundary = np.full(r.shape, 1) # defines the abs value of coordinates |x| or |y|
      boundary_check = np.greater(np.absolute(r), boundary) # returns True if any module of coordinate > 1
      boundary_check = -2 * boundary_check.astype(int) + 1 # True/False to -1/1
      v = np.multiply(boundary_check, v) # changes the sign of v components for where boundary_check is True
def update_with_boundaries(r, v, time_step):
       "" Time propagation with boundary conditions.
      r, v = time_propagation(r, v, time_step)
      r, v = boundary_collisions(r, v)
      return r, v
simulation2 = simulate(
      r0, v0, time_step, time_steps=400,
      update_function=update_with_boundaries)
animate particles(simulation2[0], time step, 'cal step3 movie2.mp4')
# %%
# 5. Particle-prticle collision
# -- loops over all pairs of particles,
# -- check if they are closer to each other than 2*radius and are travelling towards each other.
\# -- If yes then modifyy their velocity vectors according to a momentum conserving collision.
radius = 0.05
def particle_collisions(r, v, radius):
      dist = distance.pdist(np.swapaxes(r, 0,1), metric='euclidean') # returns condensed vector of distances
      dist = distance.squareform(dist) # find pairs closer to each other than 2*radius
      check_dist = np.less(dist, 2*radius)
      np.fill_diagonal(check_dist, False) # fill diagonal
      check_dist = np.triu(check_dist) # remove duplicates due to symmetry
      indx = np.asarray(np.where(check_dist==True))  # return indices of pairs where distance is smaller than 2*r
      indx = np.split(indx, indx.shape[1], axis=1)
      for i in indx: # i - pair of indexes of the colliding particles
             # Please add the check on the velocities that ensures that the
             # particles are travelling towards each other.
             # Please work with vector expressions instead of treating the x and y
             # components separately
             \# v_i = v[:, i[0]] \# i-particle pair of v components (x, y)
             \# v_j = v[:, i[1]] \# j-particle pair of v components (x, y)
            v_jx = v[:, i[1]][0] + (v[:, i[0]][0] - v[:, i[1]][0])
            v_j_y = v[:, i[1]][1] + (v[:, i[0]][1] - v[:, i[1]][1])
            v[:, i[0]] = [v_i_x, v_i_y]
            v[:, i[1]] = [v_j_x, v_j_y]
      return r, v
def update_with_interactions_slow(r, v, time_step, radius):
      r, v = time_propagation(r, v, time_step)
      r, v = boundary_collisions(r, v)
      r, v = particle_collisions(r, v, radius)
      return r, v
simulation3 = simulate(
      r0, v0, time_step, time_steps=100,
      update_function=update_with_interactions_slow, radius=radius)
animate_particles(simulation3[0], time_step, 'cal_step3_movie3.mp4')
# %%
# 6. Fast hard spheres
# Use the KDTree class in scipy.spatial to rapidly find all pairs of particles
# closer than two times the radius to each other.
def particle collisions fast (r, v, radius):
      tree = cKDTree(np.swapaxes(r, 0, 1))
      \verb|dist_set| = \verb|tree.query_pairs(2*radius)| # method; returns Python SET of pairs of points whose distance is at most relation to the state of the
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for i in list(dist_set): # i - pair of indexes of the colliding particles
                # Please consider all the comments in the function 'particle_collisions' above
                \# v_i = v[:, i[0]] \# i-particle pair of v components (x, y)
               \# v_j = v[:, i[1]] \# j-particle pair of v components (x, y)
               v_i = v[:, i[0]][0] - (v[:, i[0]][0] - v[:, i[1]][0])
               v_iy = v[:, i[0]][1] - (v[:, i[0]][1] - v[:, i[1]][1])
               v_jx = v[:, i[1]][0] + (v[:, i[0]][0] - v[:, i[1]][0])

v_jy = v[:, i[1]][1] + (v[:, i[0]][1] - v[:, i[1]][1])
               v[:, i[0]] = [v_i_x, v_i_y]
               v[:, i[1]] = [v_j_x, v_j_y]
        return r, v
 def update with interactions fast (r, v, time step, radius):
        r, v = time_propagation(r, v, time_step)
        r, v = boundary\_collisions(r, v)
        r, v = particle_collisions_fast(r, v, radius)
        return r. v
 simulation4 = simulate(
        r0, v0, time_step, time_steps=1000,
        update_function=update_with_interactions_fast, radius=radius)
 animate_particles(simulation4[0], time_step, 'ca1_step3_movie4.mp4')
 # %%
 # 7. Statistical analysis of interacting particles
 plot_statistical_analysis(simulation4[0], simulation4[1], time_step, 'cal_step3_figure_summary.svg')
diff --git a/ca1_step_2_solution.py b/ca1_step_2_solution.py
index 446ebel..ebfaadd 100644
--- a/ca1_step_2_solution.py
+++ b/ca1_step_2_solution.py
@@ -1,219 +1,246 @@
+# Please use a docstring here not comments
           CA 1: STEP 2
 import os
 import itertools
 import h5py
 from scipy import stats from scipy.spatial import distance
 import numpy as np
 from matplotlib import pyplot as plt
 import matplotlib.animation as animation
 def plot_statistical_analysis(R, V, time_step, filename):
        Analyzes R and V and plots statistical analysis results
        time = np.arange(0, R.shape[0]*time_step, time_step) # [sec]
        # General comment on reshaping of arrays
        # Please do not hard code the dimensions in the reshaping of arrays
        # This makes your function only work for 1000 timesteps with 400 particles
        # Instead of doing reshaping with given shapes please use the ndarray method .flatten()
         # It is also convenient to refrain from defining flattened variables and instead
        # flatten the array when calling .fit and .hist, i.e.
        # plt.hist(R[:, 0, :].flatten(), ...)
         # Prepare data for x,y velocities and coordinates plots:
        # 1. x and y positions:
        R_{reshaped_x} = np.reshape(R[:,0,:], (400*1000,))
        R_reshaped_y = np.reshape(R[:,1,:], (400*1000,))
V_reshaped_x = np.reshape(V[:,0,:], (400*1000,))
        V_{reshaped_y} = np.reshape(V[:,1,:], (400*1000,))
        v x loc, v x scale = stats.norm.fit(V reshaped x) # fit model to the data
         v_x = pdf = stats.norm.pdf (np.arange(-0.5, 0.5, 0.01), loc=v_x = loc, scale=v_x = cale) \# modelled distribution (np.arange(-0.5, 0.5, 0.01), loc=v_x = loc, scale=v_x = cale) \# modelled distribution (np.arange(-0.5, 0.5, 0.01), loc=v_x = loc, scale=v_x = cale) # modelled distribution (np.arange(-0.5, 0.5, 0.01), loc=v_x = loc, scale=v_x = cale) # modelled distribution (np.arange(-0.5, 0.5, 0.01), loc=v_x = loc, scale=v_x = cale) # modelled distribution (np.arange(-0.5, 0.5, 0.01), loc=v_x = cale) # modelled distribution (np.arange(-0.5, 0.5, 0.01), loc=v_x = cale) # modelled distribution (np.arange(-0.5, 0.5, 0.01), loc=v_x = cale) # modelled distribution (np.arange(-0.5, 0.01), loc=v_x = c
        v_y_loc, v_y_scale = stats.norm.fit(V_reshaped_y)
         v\_y\_pdf = stats.norm.pdf(np.arange(-0.5, 0.5, 0.01), loc=v\_y\_loc, scale=v\_y\_scale) 
        x loc, x scale = stats.uniform.fit(R reshaped x)
        x_pdf = stats.uniform.pdf(np.arange(-1.5, 1.5, 0.01), loc=x_loc, scale=x_scale)
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y_loc, y_scale = stats.uniform.fit(R_reshaped_y)
y_pdf = stats.uniform.pdf(np.arange(-1.5, 1.5, 0.01), loc=y_loc, scale=y_scale)
plt.subplots_adjust(left=None, bottom=None, right=None, top=None,
                    wspace=0.5, hspace=0.7) # Tune the subplot layout
# Prepare data for pair distance distribution function:
# Please use NumPy broadcasting to implement thie pair distances
# The goal is use the basic mechanism of broadcasting (not scipy.spatial.distance.pdist)
R_split_time = np.split(R, R.shape[0], axis=0)
def calculate_distances(R_split_time):
   dist = distance.pdist(np.swapaxes(np.squeeze(R_split_time), 0, 1),
                      metric='euclidean') # returns condensed vector of distances
   return dist
Dist = list(map(calculate_distances, R_split_time))  # produces the list of condensed distances
Dist = np.asarray(Dist) # (1000, 79800) ndarray
# Prepare data for velocity norm pdf:
V_norm = np.linalg.norm(V, ord=2, axis=1)
V_reshaped = np.reshape(V_norm, (400*1000,))
v_loc, v_scale = stats.rayleigh.fit(V_reshaped)
v_pdf = stats.rayleigh.pdf(np.arange(0, 0.5, 0.01), loc=v_loc, scale=v_scale)
# Prepare data for E_k and V_tot:
V_sum = np.sum(V, 2)
V_tot = np.linalg.norm(V_sum, ord=2, axis=1)
# This is the second time the function computes V_norm
# Please remove one of the calculations
V_norm = np.linalg.norm(V, ord=2, axis=1)
V_norm.shape
# This row of code is not doing anything please clean up
V norm.shape
E_k = np.sum(np.square(V_norm)/2, 1)
# PLOTS:
plt.figure(1, figsize=(20, 30)) # â\200\231figsize â\200\231 to make the figure larger
plt.subplots_adjust(left=None, bottom=None, right=None, top=None,
                   wspace=0.5, hspace=1.5) # Tune the subplot layout
# V tot plot:
plt.subplot(4,2,1)
plt.plot(time, V_tot, color='blue', linewidth=1.5)
plt.xlabel('Time')
plt.ylabel('Total velocity')
plt.grid(True)
# E_k plot:
plt.subplot(4,2,2)
plt.plot(time, E_k, color='blue', linewidth=1.5)
plt.ylim(0, 5) # can be removed to look at the noisy component
plt.xlabel('Time')
plt.ylabel('Kinetic energy')
plt.grid(True)
# pair distance distribution function plot:
plt.hist(Dist[500], 70, density=True, facecolor='blue', alpha=0.75) # plot hist only at a single time-point
plt.xlabel('Pair distance, $d_{ij}$')
plt.ylabel('Probability')
plt.xlim(0, 3.0) # can be removed to look at the noisy component
plt.grid(True)
# Plot velocity norm pdf and hist:
plt.subplot(4,2,4)
plt.hist(V_reshaped, 70, density=True, facecolor='blue', alpha=0.75)
plt.plot(np.arange(0, 0.5, 0.01), v_pdf, linewidth=2, color='red')
plt.xlabel('Velocity norm')
plt.ylabel('Probability')
plt.xlim(0, 0.5) # can be removed to look at the noisy component
plt.grid(True)
# Plots for x,y velocities and coordinates:
plt.subplot(4,2,5) # Position number option
plt.hist(V_reshaped_x, 40, density=True, facecolor='blue', alpha=0.75)
\verb|plt.plot(np.arange(-0.5, 0.5, 0.01), v_x_pdf, linewidth=2, color='red')| \\
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plt.xlabel('x velocities')
    plt.ylabel('Probability')
    plt.xlim(-0.4, 0.4) \# can be removed to look at the noisy component
    plt.grid(True)
    plt.subplot(4,2,6) # Position number option
    plt.hist(V_reshaped_x, 40, density=True, facecolor='blue', alpha=0.75)
    plt.plot(np.arange(-0.5, 0.5, 0.01), v_x_pdf, linewidth=2, color='red')
    plt.xlabel('y velocities')
    plt.ylabel('Probability')
    plt.xlim(-0.4, 0.4) # can be removed to look at the noisy component
    plt.grid(True)
    plt.subplot(4,2,7)
    plt.hist(R_reshaped_x, 70, density=True, facecolor='blue', alpha=0.75)
    plt.plot(np.arange(-1.5, 1.5, 0.01), x_pdf, linewidth=2, color='red')
    plt.xlabel('x coordinates')
    plt.ylabel('Probability')
    plt.xlim(-1.5, 1.5) \# can be removed to look at the noisy component
    plt.grid(True)
    plt.subplot(4,2,8)
    plt.hist(R_reshaped_y, 70, density=True, facecolor='blue', alpha=0.75) plt.plot(np.arange(-1.5, 1.5, 0.01), y_pdf, linewidth=2, color='red')
    plt.xlabel('y coordinates')
    plt.ylabel('Probability')
    plt.xlim(-1.5, 1.5) # can be removed to look at the noisy component
    plt.grid(True)
    plt.savefig(filename)
    plt.show()
def animate_particles(R, time_step, filename='movie.mp4'):
    """Generates an mp4 movie with the particle trajectories
    using MatPlotLib. """
    if os.path.isfile(filename):
        print('WARNING (animate_particles): The output file', filename, 'exists. Skipping animation.')
        return
    frames = R.shape[0]
    frames_per_second = 1. / time_step
    fig = plt.figure(figsize=(6, 6))
    ax = plt.subplot()
    plt.xlabel('x')
    plt.ylabel('y')
    plt.axis('image')
    plt.xlim([-1.0, 1.0])
    plt.ylim([-1.0, 1.0])
    markers = ax.scatter([], [], facecolor='red', s=180, alpha=0.5)
text = plt.text(-0.9, 0.9, 'frame =', ha='left')
    def update(frame, markers, text):
        print('frame =', frame)
        r = R[frame]
        markers.set_offsets(r.T)
        text.set_text('frame = {}'.format(frame))
    anim = animation.FuncAnimation(
        fig, update,
        frames=frames, interval=50,
        fargs=(markers, text),
        blit=False, repeat=False,
    writer = animation.writers['ffmpeg'](fps=frames_per_second)
    anim.save(filename, writer=writer, dpi=100)
    plt.close()
if __name__ == '__main__':
    # Read h5:
    with h5py.File('cal_step1_output_data.h5', 'r') as hfile:
        print(hfile.keys())
        dataset_R = hfile['R']
        R = dataset R[:]
        dataset_V = hfile['V']
        V = dataset_V[:]
        time_step_s = dataset_V.attrs['time_step_s']
        radius = dataset_V.attrs['radius']
        v_variance = dataset_V.attrs['v_variance']
    # Calling del on hfile is not required since you used the with statement
```