KyotoUx-009x (/github/ryo0921/KyotoUx-009x/tree/master) / 05 (/github/ryo0921/KyotoUx-009x/tree/master/05)

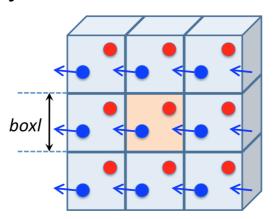
# **Stochastic Processes: Data Analysis and Computer Simulation**

### **Brownian motion 3: data analysis**

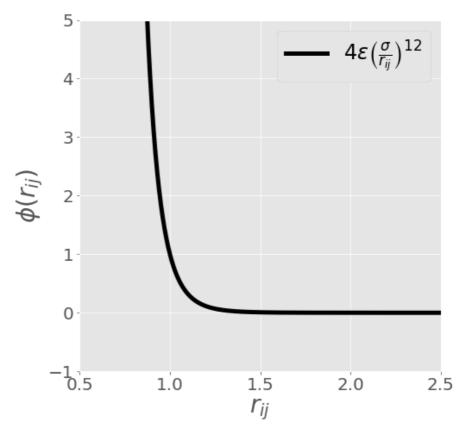
### 3. Interacting Brownian particles

# 3.1. Necessary changes for interacting Brownian particles

### **Periodic boundary conditions**



### Inter-particle interaction



$$U = \sum_{r_{ij}<2\sigma} \phi(r_{ij}) = \sum_{r_{ij}<2\sigma} 4\epsilon \left(\frac{\sigma}{r_{ij}}\right)^{12}, \quad \mathbf{F}_i = -\frac{\partial U}{\partial \mathbf{r}_{ij}} = \sum_{r_{ij}<2\sigma} 48\epsilon \left(\frac{\sigma}{r_{ij}}\right)^{12} \frac{\mathbf{r}_{ij}}{r_{ij}^2} \quad (\text{I}1, \text{L}2)$$

$$\mathbf{r}_{ij} = \mathbf{r}_j - \mathbf{r}_i, \quad r_{ij} = |\mathbf{r}_{ij}|$$

# 3.2. Simulation code for interacting Brownian particles with animation

### Import libraries

In [1]: % matplotlib nbagg
import numpy as np # import numpy library as np
import matplotlib.pyplot as plt # import pyplot library as plt
import matplotlib.animation as animation # import animation modules from
from mpl\_toolkits.mplot3d import Axes3D # import Axes3D from mpl\_toolk
from numpy import newaxis # import newaxis used for inter-particle for
plt.style.use('ggplot') # use "ggplot" style for graphs

### Define init function for FuncAnimation

```
In [2]: def init():
            global R,V,W,F,Rs,Vs,Ws,time # define global variables
            V[:,:] = 0.0
                               # create random particle configuration without c
                               # initialize all the variables to zero
            W[:,:] = 0.0

F[:,:] = 0.0
                               # initialize all the variables to zero
                               # initialize all the variables to zero
            Rs[:,:,:] = 0.0 # initialize all the variables to zero
            Rs[0,:,:] = R[:,:] # store initial particle positions in Rs
            Vs[:,:,:] = 0.0 # initialize all the variables to zero
            Ws[:,:,:] = 0.0
                              # initialize all the variables to zero
                                # initialize all the variables to zero
            time[:] = 0.0
            title.set text(r'') # empty title
            line.set data([],[]) # set line data to show the trajectory of parti
            line.set 3d properties([]) # add z-data separately for 3d plot
            particles.set_data([],[]) # set position current (x,y) position data
            particles.set 3d properties([]) # add current z data of particles to
            return particles, title, line # return listed objects that will be dra
```

#### Define animate function for FuncAnimation

The Euler method used in previous animate function is replaced with the Leap-Frog method in the following new animate function because of the numerical instability taking place with the former method for the present case of interacting Brownian particles.

```
In [3]: def animate(i):
            global R,V,W,F,Rs,Vs,Ws,time # define global variables
            time[i]=i*dt # store time in each step in an array time
            particleforces() # compute inter-particle force F by examining all n
            W = std*np.random.randn(nump,dim) # generate an array of random forc
            V = V*(1-zeta/m*dt)+F/m*dt+W/m \# update velocity via Eq.(I11) using
            V = (V*(1-zeta/2/m*dt)+F/m*dt+W/m)/(1+zeta/2/m*dt) # update velocity
            R = R+V*dt # update position via Eq.(I12) using the L-F method
            Rs[i,:,:]=R # accumulate particle positions at each step in an array
            Vs[i,:,:]=V # accumulate particle velocitys at each step in an array
            Ws[i,:,:]=W # accumulate random forces at each step in an array Ws
            title.set\_text(r"t = "+str(time[i])+"/"+str((nums-1)*dt))  # set the
            line.set_data(Rs[:i+1,n,0],Rs[:i+1,n,1]) # set the line in 2D (x,y)
            line.set_3d_properties(Rs[:i+1,n,2]) # add z axis to set the line in
            particles.set_data(pbc(R[:,0],box[0]), pbc(R[:,1],box[1])) # set the
            particles.set_3d_properties(pbc(R[:,2],box[2])) # add z axis to set
            return particles, title, line # return listed objects that will be dra
```

## Newly defined functions for interacting Brownian particles

```
In [4]: def pbc(r, lbox): # enforce Periodic Boundary Conditions for all positic
            return np.fmod(r+lbox,lbox)
        def distance(r1,r2,lbox): # Compute distance vector R2 - R1 with PBC
            return r2-r1-np.around((r2-r1)/lbox)*lbox
        def fij(r2,rij): # calculate Fij=dU/drij
             f=-24*eps*(2*(r2/sig**2)**(-6)-(r2/sig**2)**(-3))/r2*rij # Lennard-
            f=-48*eps*((r2/sig**2)**(-6))/r2*rij # soft-core potential
        def particleforces(): # compute inter-particle force F by examining all
            global F
            F[:,:] = 0.0
            for n in range(nump): # repeat below for all particles
                rij = distance(R[n,:], R, box) # distance vectors rij=R i-R j fc
                r2 = np.linalg.norm(rij, axis=1)**2 # square distance rij**2
                nei = (r2 < (2.0*sig)**2) # list neighbor particles of j
                nei[n] = False # ignore self pair (i=j)
                F[n,:] = np.sum(fij(r2[nei, newaxis], rij[nei,:]), axis=0) # tot
        def initconf(): # create random particle configuration without overlapp
            global R,V,W,F,Rs,Vs,Ws,time
            for n in range(nump): # repeat below from n=0 to nump-1
                nn=0 # set overlap true to perform while loop below for the n-th
                while nn == 0: # repeat the loop below while overlap is true (nn
                    R[n,:]=np.random.rand(dim)*box # generate a position candida
                    nn = 1 # initialize overlap as false
                    for 1 in range(n): # examine overlap generated positions (fr
                        rij = distance(R[n,:],R[1,:],box) # calculate distance v
                        r2 = np.linalg.norm(rij)**2 # calculate the squared dis
                        if r2 < (0.90*sig)**2: # check if the distance is smalle</pre>
                        \# Yes -> perform below (nn=0) -> repeat while loop, No (
                            nn = 0 # set overlap true
```

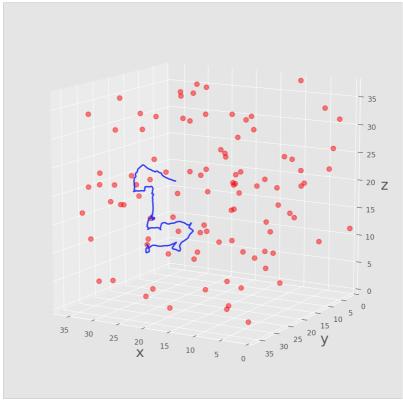
### Set parameters and initialize variables

```
In [5]: dim = 3
                    # system dimension (x,y,z)
        nump = 100 # number of interacting Brownian particles to simulate
        nums = 4096 # number of simulation steps
           = 0.01 # set time increment, \Delta t
        zeta = 1.0 # set friction constant, \zeta
            = 1.0 # set particle mass, m
        kBT = 1.0 # set thermal energy, k\_B T
        std = np.sqrt(2*kBT*zeta*dt) # calculate std for \Delta W via Eq.(F11)
        sig = 1.0 # unit of length of inter-particle potential
        eps = 1.0 # unit of energy inter-particle potential
            = 0.001 ##volume fraction of particles < 0.45
        box1 = np.power(nump*np.pi/6/vf,1/3) # calculate the side length of unit
        print('Volume fraction =',vf,' boxl =',boxl) # print vf and boxl
        box = np.array([boxl,boxl,boxl])*sig # set array box[dim]
        np.random.seed(0) # initialize random number generator with a seed=0
        R = np.zeros([nump,dim]) # array to store current positions and set init
        V = np.zeros([nump,dim]) # array to store current velocities and set ini
        W = np.zeros([nump,dim]) # array to store current random forcces
        F = np.zeros([nump,dim]) # rray to store current particle orcces
        Rs = np.zeros([nums,nump,dim]) # array to store positions at all steps
        Vs = np.zeros([nums,nump,dim]) # array to store velocities at all steps
        Ws = np.zeros([nums,nump,dim]) # array to store random forces at all ste
        time = np.zeros([nums]) # an array to store time at all steps
```

Volume fraction = 0.001 box1 = 37.4110192682

### Perform and animate the simulation using FuncAnimation

```
In [6]: fig = plt.figure(figsize=(10,10)) # set fig with its size 10 x 10 inch
        ax = fig.add subplot(111,projection='3d') # creates an additional axis t
        ax.set xlim(0.0,box[0]) # set x-range
        ax.set_ylim(0.0,box[1]) # set y-range
        ax.set zlim(0.0,box[2]) # set z-range
        ax.set xlabel(r"x",fontsize=20) # set x-lavel
        ax.set_ylabel(r"y",fontsize=20) # set y-lavel
        ax.set_zlabel(r"z",fontsize=20) # set z-lavel
        ax.view init(elev=12,azim=120) # set view point
        particles, = ax.plot([],[],[],linestyle='None',color='r',marker='o',ms=2
        title = ax.text(0.,0.,0.,r'',transform=ax.transAxes,va='center') # defin
        line, = ax.plot([],[],[],'b',lw=2,alpha=0.8) # define object line
        n = 0 # trajectry line is plotted for the n-th particle
        anim = animation.FuncAnimation(fig,func=animate,init func=init,
                    frames=nums,interval=5,blit=True,repeat=False)
        ## If you have ffmpeg installed on your machine
        ## you can save the animation by uncomment the last line
        ## You may install ffmpeg by typing the following command in command prc
        ## conda install -c menpo ffmpeg
        #anim.save('movie.mp4',fps=20,dpi=400)
```



t = 40.95/40.95

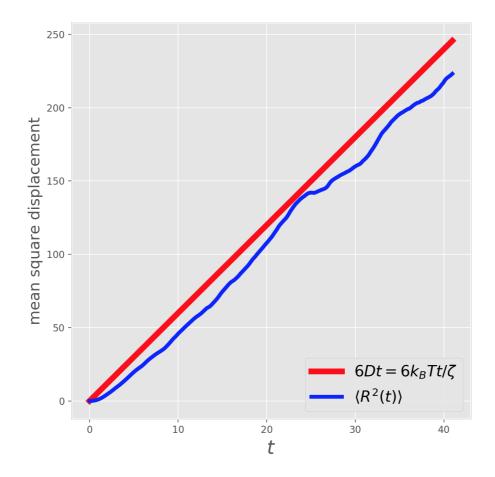
## 3.3. The mean square displacement and the diffusion constant

```
In [7]:
        # mean square displacement vs time
        msd = np.zeros([nums])
        for i in range(nums):
            for n in range(nump):
                msd[i]=msd[i]+np.linalg.norm(Rs[i,n,:]-Rs[0,n,:])**2 # (R(t) - F)
            msd[i] = msd[i]/nump
        dmsd = np.trapz(msd, dx=dt)/(3*(nums*dt)**2)
        print('D_0 =',kBT/zeta,'(Theoretical)')
        print('D =',dmsd,'(Simulation via MSD)')
        print('Volume fraction =',vf) ### print vf
        fig, ax = plt.subplots(figsize=(7.5,7.5))
        ax.set xlabel(r"$t$", fontsize=20)
        ax.set ylabel(r"mean square displacement", fontsize=16)
        ax.plot(time,6*kBT/zeta*time,'r',lw=6,label=r'$6Dt={6k BT t}/{\zeta}$')
        ax.plot(time,msd,'b',lw=4, label=r'$\langle R^2(t)\rangle$')
        ax.legend(fontsize=16, loc=4)
        plt.show()
```

```
D_0 = 1.0 (Theoretical)

D = 0.885382884029 (Simulation via MSD)

Volume fraction = 0.001
```



### References

- The SciPy.org website, <a href="https://docs.scipy.org/doc/numpy/user/basics.broadcasting.html">https://docs.scipy.org/doc/numpy/user/basics.broadcasting.html</a> (https://docs.scipy.org/doc/numpy/user/basics.broadcasting.html)
- Listed in the major unsolved problems in physics, https://en.wikipedia.org /wiki/List of unsolved problems in physics#Condensed matter physics (https://en.wikipedia.org

/wiki/List of unsolved problems in physics#Condensed matter physics)

### Summary of simulation method

### **Original differential equation**

i: particle index

$$\frac{d\mathbf{R}_i(t)}{dt} = \mathbf{V}_i(t) \tag{13}$$

$$\frac{d\mathbf{R}_{i}(t)}{dt} = \mathbf{V}_{i}(t)$$

$$m\frac{d\mathbf{V}_{i}(t)}{dt} = -\zeta \mathbf{V}_{i}(t) + \mathbf{F}_{i}(t) + \mathbf{F}_{R}(t)$$
(I3)

with 
$$\langle \mathbf{F}_R(t) \rangle = \mathbf{0}$$
 (I5)

$$\langle \mathbf{F}_R(t)\mathbf{F}_R(0)\rangle = 2k_B T \zeta \mathbf{I}\delta(t) \tag{I6}$$

#### **Euler method**

i: index of particles, n: time step

$$\mathbf{V}_{i,n+1} = \left(1 - \frac{\zeta}{m} \Delta t\right) \mathbf{V}_{i,n} + \frac{1}{m} \mathbf{F}_{i,n} \Delta t + \frac{1}{m} \Delta \mathbf{W}_n \tag{17}$$

$$\mathbf{R}_{i\,n+1} = \mathbf{R}_{i\,n} + \mathbf{V}_{i\,n}\Delta t \tag{I8}$$

with 
$$\langle \Delta \mathbf{W}_n \rangle = \mathbf{0}$$
 (I9)

$$\langle \Delta \mathbf{W}_n \Delta \mathbf{W}_l \rangle = 2k_B T \zeta \Delta t \mathbf{I} \delta_{nl} \tag{I10}$$

#### Leap-Frog method

i: index of particles, n: time step

$$\mathbf{V}_{i,n+\frac{1}{2}} = \mathbf{V}_{i,n-\frac{1}{2}} - \frac{\zeta}{m} \Delta t \mathbf{V}_{i,n} + \frac{1}{m} \mathbf{F}_{i,n} \Delta t + \frac{1}{m} \Delta \mathbf{W}_{n}$$

$$= \mathbf{V}_{i,n-\frac{1}{2}} - \frac{\zeta}{m} \Delta t \frac{1}{2} \left( \mathbf{V}_{i,n+\frac{1}{2}} + \mathbf{V}_{i,n-\frac{1}{2}} \right) + \frac{1}{m} \mathbf{F}_{i,n} \Delta t + \frac{1}{m} \Delta \mathbf{W}_{n}$$

$$\therefore \mathbf{V}_{i,n+\frac{1}{2}} = \left( \left( 1 - \frac{\zeta}{2m} \Delta t \right) \mathbf{V}_{i,n-\frac{1}{2}} + \frac{1}{m} \mathbf{F}_{i,n} \Delta t + \frac{1}{m} \Delta \mathbf{W}_{n} \right) \left( 1 + \frac{\zeta}{2m} \Delta t \right)^{-1}$$

$$\mathbf{R}_{i,n+1} = \mathbf{R}_{i,n} + \mathbf{V}_{i,n+\frac{1}{2}} \Delta t$$
(I12)

with 
$$\langle \Delta \mathbf{W}_n \rangle = \mathbf{0}$$
 (I13)

$$\langle \Delta \mathbf{W}_n \Delta \mathbf{W}_l \rangle = 2k_B T \zeta \Delta t \mathbf{I} \delta_{nl} \tag{I14}$$

### 2nd order Runge-Kutta method

i: index of particles, n: time step

$$\mathbf{V}'_{i,n+\frac{1}{2}} = \mathbf{V}_{i,n} - \frac{\zeta}{m} \frac{\Delta t}{2} \mathbf{V}_{i,n} + \frac{1}{m} \mathbf{F}_{i,n} \frac{\Delta t}{2} = \left(1 - \frac{\zeta}{m} \frac{\Delta t}{2}\right) \mathbf{V}_{i,n} + \frac{1}{m} \mathbf{F}_{i,n} \frac{\Delta t}{2}$$
(I15)

$$\mathbf{R}'_{i,n+\frac{1}{2}} = \mathbf{R}_{i,n} + \mathbf{V}_{i,n} \frac{\Delta t}{2} \tag{I16}$$

$$\mathbf{R}'_{i,n+\frac{1}{2}} = \mathbf{R}_{i,n} + \mathbf{V}_{i,n} \frac{\Delta t}{2}$$

$$\mathbf{V}_{n,i+1} = \mathbf{V}_{i,n} - \frac{\zeta}{m} \Delta t \mathbf{V}'_{n,i+\frac{1}{2}} + \frac{1}{m} \Delta t \mathbf{F}_{i,n+\frac{1}{2}} + \frac{1}{m} \Delta \mathbf{W}_{i}$$

$$\mathbf{R}_{n,i+1} = \mathbf{R}_{n,i} + \mathbf{V}'_{n,i+\frac{1}{2}} \Delta t$$
(I16)
(I17)

$$\mathbf{R}_{n,i+1} = \mathbf{R}_{n,i} + \tilde{\mathbf{V}}'_{n,i+\frac{1}{2}} \Delta t \tag{I18}$$

with 
$$\langle \Delta \mathbf{W}_n \rangle = \mathbf{0}$$
 (I19)

$$\langle \Delta \mathbf{W}_n \Delta \mathbf{W}_l \rangle = 2k_B T \zeta \Delta t \mathbf{I} \delta_{nl} \tag{I20}$$