

# Understanding Molecular Simulation 2010

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# Outline

- Statistical Mechanics I: Entropy, Boltzmann weight, partition function, canonical ensemble [Frenkel Lecture 1, Das Lecture 1]
- Statistical Mechanics II: Different ensembles, Dynamics [Frenkel Lecture 2, Sastry Tutorial Lecture 1]
- Molecular dynamics simulations - basic principles, boundary conditions, neighbor lists, truncation and tail corrections [Frenkel Lecture 3]
- This Tutorial Lecture:
  - Review basics of molecular dynamics
  - Molecular dynamics in different ensembles
  - Event driven MD
  - Molecular systems
  - Long range interactions



# Molecular Dynamics – Basic concepts

- Integrate the equations of motion to generate a trajectory.
- Average over the trajectory is the average that we seek for properties of the system.
- Statistical mechanics provides formulation to extract useful quantities.
- Both equilibrium thermodynamic and dynamical quantities are obtained.
- Structure of MD simulations: Initialization, force calculation, integration of equations of motion, sampling to calculate desired averages.

## Algorithm 3 (A Simple Molecular Dynamics Program)

program md	simple MD program
call init	initialization
t=0	
do while (t.lt.tmax)	MD loop
call force(f,en)	determine the forces
call integrate(f,en)	integrate equations of motion
t=t+delt	
call sample	sample averages
enddo	
stop	
end	

*Comment to this algorithm:*

- 1. Subroutines init, force, integrate, and sample will be described in Algorithms 4, 5, and 6, respectively. Subroutine sample is used to calculate averages like pressure or temperature.*



## Algorithm 5 (Calculation of the Forces)

subroutine force(f,en)	determine the force and energy
en=0	
do i=1,npart	
f(i)=0	set forces to zero
enddo	
do i=1,npart-1	
do j=i+1,npart	loop over all pairs
xr=x(i)-x(j)	
xr=xr-box*nint(xr/box)	periodic boundary conditions
r2=xr**2	
if (r2.lt.rc2) then	test cutoff
r2i=1/r2	
r6i=r2i**3	
ff=48*r2i*r6i*(r6i-0.5)	Lennard-Jones potential
f(i)=f(i)+ff*xr	update force
f(j)=f(j)-ff*xr	
en=en+4*r6i*(r6i-1)-ecut	update energy
endif	
enddo	
enddo	
return	
end	

# Equations of Motion

Discretize equations of motion in order to integrate.

Desired properties: Preserve symmetries of the original equations (time reversal..), conservation of energy, permit large time steps of integration...

**Verlet Algorithm:** Sum Taylor series for positive and negative time

Steps:

$$r(t + \Delta t) = r(t) + v(t)\Delta t + \frac{f(t)}{2m}\Delta t^2 + \frac{\Delta t^3}{3!}\ddot{r} + \mathcal{O}(\Delta t^4)$$

$$r(t - \Delta t) = r(t) - v(t)\Delta t + \frac{f(t)}{2m}\Delta t^2 - \frac{\Delta t^3}{3!}\ddot{r} + \mathcal{O}(\Delta t^4)$$

$$r(t + \Delta t) + r(t - \Delta t) = 2r(t) + \frac{f(t)}{m}\Delta t^2 + \mathcal{O}(\Delta t^4)$$

$$r(t + \Delta t) = 2r(t) - r(t - \Delta t) + \frac{f(t)}{m}\Delta t^2$$

Update with positions at two times. No velocity. Good accuracy.

But can define velocity in terms of adjacent time positions:

$$v(t) = \frac{r(t + \Delta t) - r(t - \Delta t)}{2\Delta t} + \mathcal{O}(\Delta t^2)$$

Verlet algorithm equivalent to Leap-Frog and Velocity Verlet:

### Leap Frog

$$r(t + \Delta t) = r(t) + \Delta t v(t + \Delta t/2)$$

$$v(t + \Delta t/2) = v(t - \Delta t/2) + \Delta t \frac{f(t)}{m}$$

### Velocity Verlet

$$r(t + \Delta t) = r(t) + \Delta t v(t) + \Delta t^2 \frac{f(t)}{2m}$$

$$v(t + \Delta t) = v(t) + \Delta t \frac{f(t) + f(t + \Delta t)}{2m}$$

How do we compare different integrators?

Long time accuracy is not a possible criterion. Lyapunov instability.

Symmetries, preservation of phase space volume, conservation of energy... efficient sampling of correct phase space.



## Algorithm 6 (Integrating the Equations of Motion)

```
subroutine integrate(f,en)
```

```
sumv=0
```

```
sumv2=0
```

```
do i=1,npart
```

```
  xx=2*x(i)-xm(i)+delt**2*f(i)
```

```
  vi=(xx-xm(i))/(2*delt)
```

```
  sumv=sumv+vi
```

```
  sumv2=sumv2+vi**2
```

```
  xm(i)=x(i)
```

```
  x(i)=xx
```

```
enddo
```

```
temp=sumv2/(3*npart)
```

```
etot=(en+0.5*sumv2)/npart
```

```
return
```

```
end
```

integrate equations of motion

MD loop

Verlet algorithm (4.2.3)

velocity (4.2.4)

velocity center of mass

total kinetic energy

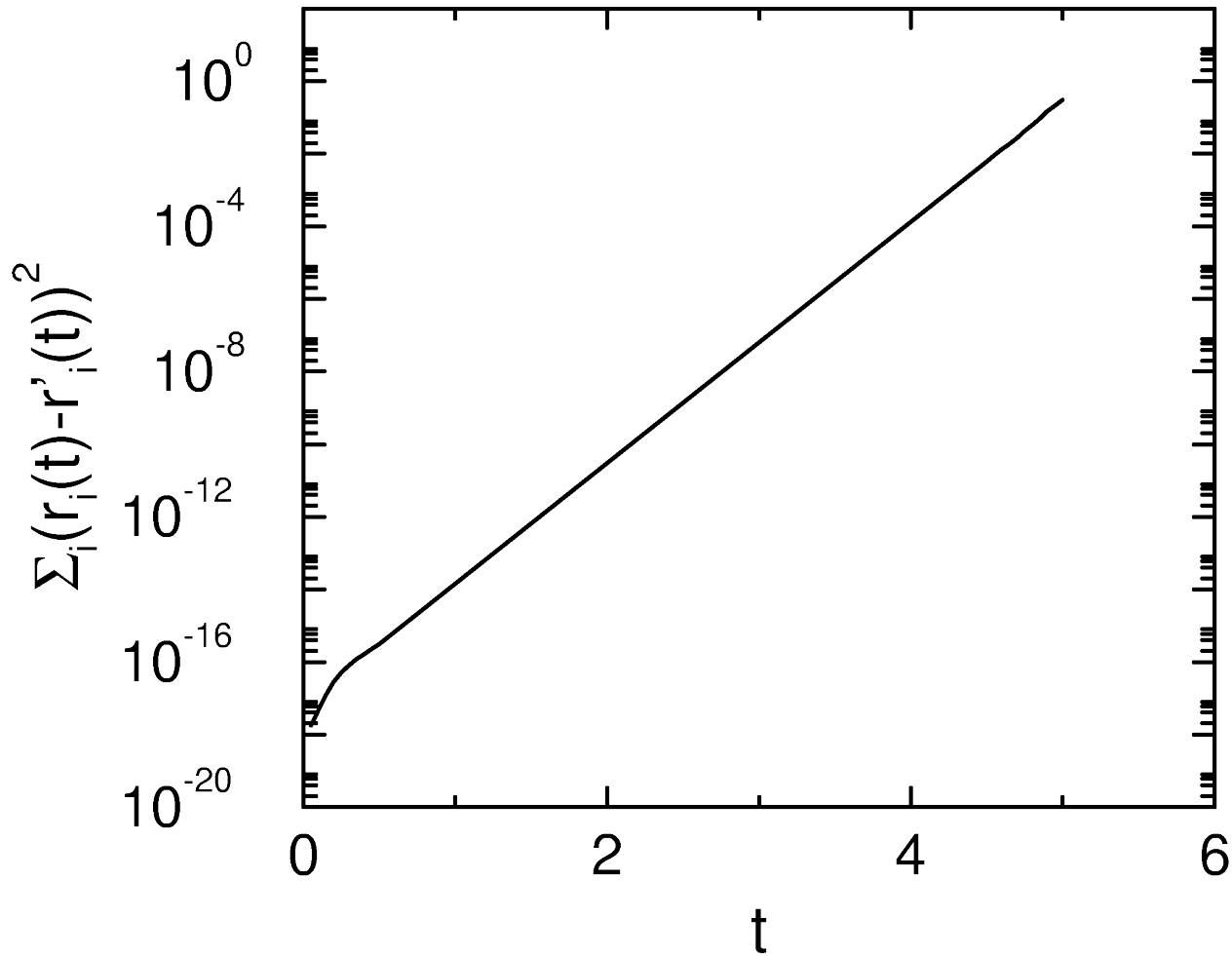
update positions previous time

update positions current time

instantaneous temperature

total energy per particle

Lyapunov Instability: Exponential divergence of nearby trajectories. True for typical many body systems.



# Time Reversible Algorithms: Liouville Formulation

Time evolution of functions of coordinates and momenta:

$$\dot{f}(\mathbf{r}, \mathbf{p}) = \dot{\mathbf{r}} \frac{\partial f}{\partial \mathbf{r}} + \dot{\mathbf{p}} \frac{\partial f}{\partial \mathbf{p}} \equiv iL f \quad \text{Liouville Operator} - iL = \dot{\mathbf{r}} \frac{\partial}{\partial \mathbf{r}} + \dot{\mathbf{p}} \frac{\partial}{\partial \mathbf{p}}$$

Formal solution:  $f(\mathbf{r}(t), \mathbf{p}(t)) = \exp(iLt) f(\mathbf{r}, \mathbf{p})$

How to develop useful integration schemes?

Consider position and momentum parts of  $L$  separately:

$$iL_r = \dot{\mathbf{r}} \frac{\partial}{\partial \mathbf{r}}; \quad iL_p \dot{\mathbf{p}} \frac{\partial}{\partial \mathbf{p}}$$

Operation by  $iL_r$  leads to shift of coordinates:  
Similarly  $iL_p$  leads to shift in momenta.

$$\begin{aligned} f(t) &= f(0) + iL_r t f(0) + \frac{(iL_r t)^2}{2} f(0) + \dots \\ &= \sum_{n=0}^{\infty} \frac{(\dot{\mathbf{r}}(0)t)^n}{n!} \frac{\partial^n}{\partial \mathbf{r}^n} f(0) \\ &= f(\mathbf{p}(0), (\mathbf{r}(0) + \dot{\mathbf{r}}(0)t). \end{aligned}$$

But since operators don't commute, we cannot write

$$\exp(iLt) = \exp(iL_r t) \times \exp(iL_p t)$$

Instead, the Trotter identity is used to write approximate expressions that are time reversible:

$$\exp(A + B) = \lim_{p \rightarrow \infty} \left[ \exp\left(\frac{A}{2p}\right) \exp\left(\frac{B}{p}\right) \exp\left(\frac{A}{2p}\right) \right]^p$$
$$\exp(A + B) = \left[ \exp\left(\frac{A}{2p}\right) \exp\left(\frac{B}{p}\right) \exp\left(\frac{A}{2p}\right) \right]^p \exp(\mathcal{O}(1/p^2))$$

We consider the case  $A/p = iL_p t/p$  and  $B/p = iL_r t/p$  and  $\Delta t = t/p$ .

One step of integration, with  $p = 2$ , then corresponds to:

$$\exp(iL_p \Delta t / 2) \exp(iL_r \Delta t) \exp(iL_p \Delta t / 2)$$

Applying step by step:

$$\exp(iL_p\Delta t/2)f(\mathbf{p}, \mathbf{r}) = f(\mathbf{p} + \frac{\Delta t}{2}\dot{\mathbf{p}}, \mathbf{r})$$

$$\exp(iL_r\Delta t)f(\mathbf{p} + \frac{\Delta t}{2}\dot{\mathbf{p}}, \mathbf{r}) = f(\mathbf{p} + \frac{\Delta t}{2}\dot{\mathbf{p}}, \mathbf{r} + \Delta t\dot{\mathbf{r}}(\Delta t/2))$$

$$\exp(iL_p\Delta t/2)f(\mathbf{p} + \frac{\Delta t}{2}\dot{\mathbf{p}}, \mathbf{r} + \Delta t\dot{\mathbf{r}}(\Delta t/2)) = f(\mathbf{p} + \frac{\Delta t}{2}\dot{\mathbf{p}} + \frac{\Delta t}{2}\dot{\mathbf{p}}(\Delta t), \mathbf{r} + \Delta t\dot{\mathbf{r}}(\Delta t/2))$$

Writing the next change in positions and momenta:

$$\mathbf{p}(0) \rightarrow \mathbf{p} + \frac{\Delta t}{2}\mathbf{F}(0) + \frac{\Delta t}{2}\mathbf{F}(\Delta t)$$

$$\mathbf{r}(0) \rightarrow \mathbf{r} + \Delta t\dot{\mathbf{r}}(\frac{\Delta t}{2})$$

$$= \mathbf{r} + \Delta t\dot{\mathbf{r}}(0) + \frac{\Delta t^2}{2m}\mathbf{F}(0).$$

This is the Velocity Verlet update!

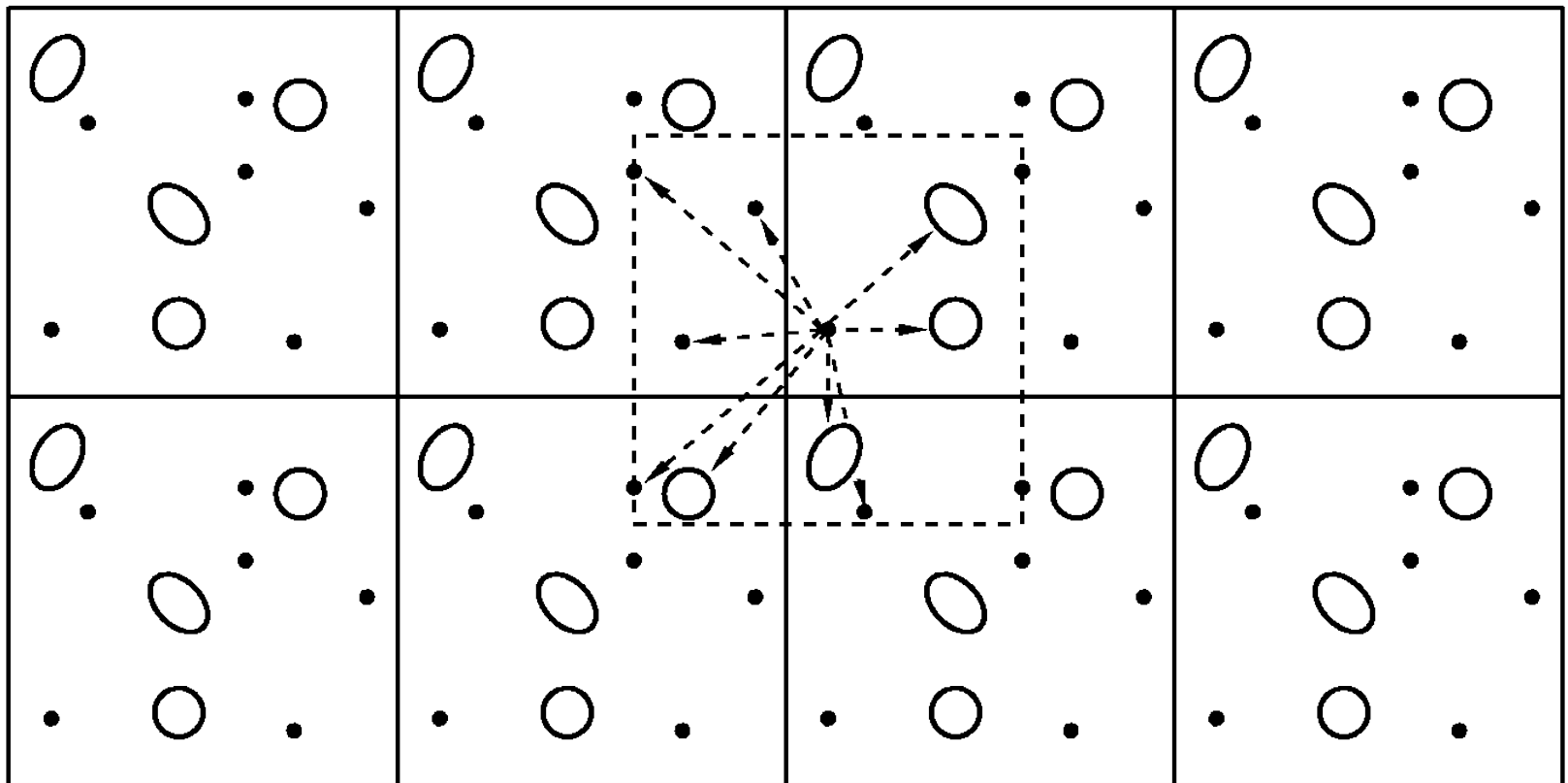
Verlet algorithm is the first member of integrators that Can be developed from this procedure.

# Periodic Boundary Conditions

Number of particles in simulations  $100 - 10^6$  but need to simulate properties of bulk systems.

Avoid free boundaries, by using periodic boundary conditions.

Calculate interactions with periodic images of particles as well.



# Truncation of the potential

Use of periodic boundary conditions not efficient if interactions with all image particles have to be calculated.

But for short ranged potentials such as the LJ potential, one truncates the potential beyond a cutoff  $r_c$ .

Thus, we use

$$U(r) = \begin{cases} 4\epsilon \left[ \left(\frac{\sigma}{r}\right)^{12} - \left(\frac{\sigma}{r}\right)^6 \right] & \text{if } r \leq r_c \\ 0 & \text{if } r \geq r_c \end{cases}$$

But long range correction needs to be calculated, or potential redefined. If we assume density of neighbors to be constant beyond cutoff:

$$U^{Corr} = \frac{N\rho}{2} \int_{r_c}^{\infty} dr U(r) 4\pi r^2 dr \qquad U^{Corr} = \frac{N8\pi\rho\epsilon\sigma^3}{3} \left[ \frac{1}{3} \left(\frac{\sigma}{r_c}\right)^9 - \left(\frac{\sigma}{r_c}\right)^3 \right]$$

For the Lennard-Jones potential

# Neighbor lists

The calculation of the energy scales as  $N^2$

But for short range interactions, finite number of neighbors for each particle, so  $O(N)$  computation should be possible.

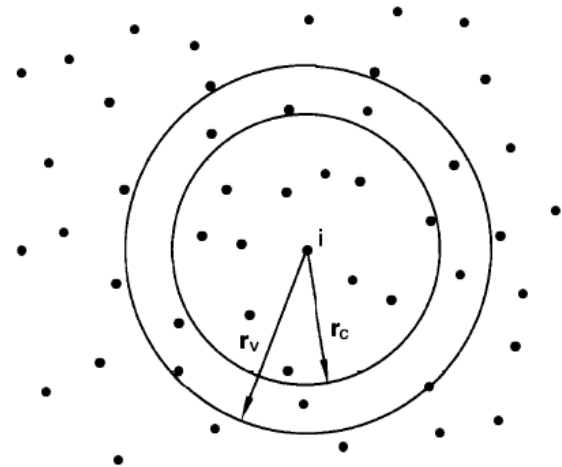
This can be done if we know the neighbors of each particle but this keeps changing.

Solution: Calculate neighbors for radius  $r_v > r_c$

Monitor how far particles move.

When maximum displacement exceed  $(r_v - r_c)/2$ , update neighbor list.

Efficient scheme – Verlet Neighbor Lists.





## Algorithm 34 (Making a Verlet List)

```
SUBROUTINE new_vlist                                makes a new Verlet list
do i=1,npart                                       initialize list
  nlist(i)=0
  xv(i)=x(i)                                       store position of particles
enddo
do i=1,npart-1
  do j=i+1,npart
    xr=x(i)-x(j)
    if (xr.gt.hbox) then                            nearest image
      xr=xr-hbox
    else if (xr.lt.-hbox) then
      xr=xr+hbox
    endif
    if (abs(xr).lt.rv) then                          add to the lists
      nlist(i)=nlist(i)+1
      nlist(j)=nlist(j)+1
      list(i,nlist(i))=j
      list(j,nlist(j))=i
    endif
  enddo
enddo
return
end
```

# Molecular Dynamics simulations in different ensembles

So far: molecular dynamics in the microcanonical ensemble (constant  $N, V, E$ ) have been discussed.

## Why do we need to consider other ensembles?

- To do single phase simulations in the presence of a first order transition – e. g., liquid - gas phase transition --  $(N, P, T)$  ensemble preferable to  $(N, V, T)$
- To simulate efficiently in conditions close to experiment – e. g., gas adsorption in porous adsorbent –  $(\mu, V, T)$  preferable to  $(N, V, T)$
- Determination of fluctuations – e. g. isothermal compressibility from volume fluctuations --  $(N, P, T)$  ensemble preferable to  $(N, V, T)$ .
- “Unphysical ensembles” for special purposes – e. g., phase equilibria.

# Molecular dynamics in the canonical (N, V, T) ensemble

In the canonical ensemble, the system is characterized by a fixed temperature, which is usually measured in MD simulations by the average kinetic energy.

One possible way of then achieving constant temperature is to constrain the kinetic energy by modifying the equations of motion:

$$\dot{r} = \frac{p}{m}; \quad \dot{p} = f - \xi(p) p$$

The constraint parameter  $\xi$  is chosen such that:

$$\frac{1}{2m} \sum_i p_i^2 = \text{constant} \longrightarrow \xi = \frac{\sum_i p_i \cdot f_i}{\sum_i p_i^2}$$

This “friction” is calculated at each integration time step.

Such modified dynamics does not generate the true canonical ensemble.

# Andersen thermostat

Another possible simple thermostat is the “Andersen thermostat” wherein the coupling with the heat bath is represented by stochastic forces that act on randomly selected particles, such that the particle acquires a new velocity drawn from the Maxwell distribution.

One must specify a “collision frequency” that determines how often such a velocity change is made.

A simple method which generates the canonical ensemble, with static properties that do not depend on collision frequency.

However, not so good for dynamics, which depend on the arbitrary collision frequency that is specified.

## Algorithm 14 (Molecular Dynamics: Andersen Thermostat)

```
program md_Andersen
call init(temp)
call force(f,en)
t=0
do while (t.lt.tmax)
  call integrate(1,f,en,temp)
  call force(f,en)
  call integrate(2,f,en,temp)
  t=t+dt
  call sample
enddo
stop
end
```

MD at constant temperature  
initialization  
determine the forces

MD loop  
first part of the eqs. of motion  
determine the forces  
second part of eqs. of motion  
sample averages

## Algorithm 15 (Equations of Motion: Andersen Thermostat)

```
subroutine integrate (switch, f
                    , en, temp)
if (switch.eq.1) then
  do i=1, npart
    x(i)=x(i)+dt*v(i)+
+    dt*dt*f(i)/2
    v(i)=v(i)+dt*f(i)/2
  enddo
else if (switch.eq.2) then
  tempa=0
  do i=1, npart
    v(i)=v(i)+dt*f(i)/2
    tempa=tempa+v(i)**2
  enddo
  tempa=tempa/(s*npart)
  sigma=sqrt(temp)
  do i=1, npart
    if (ranf().lt.nu*dt) then
      v(i)=gauss(sigma)
    endif
  enddo
endif
return
end
```

integrate equations of motion:  
with Andersen thermostat  
first step velocity Verlet

update positions current time

first update velocity

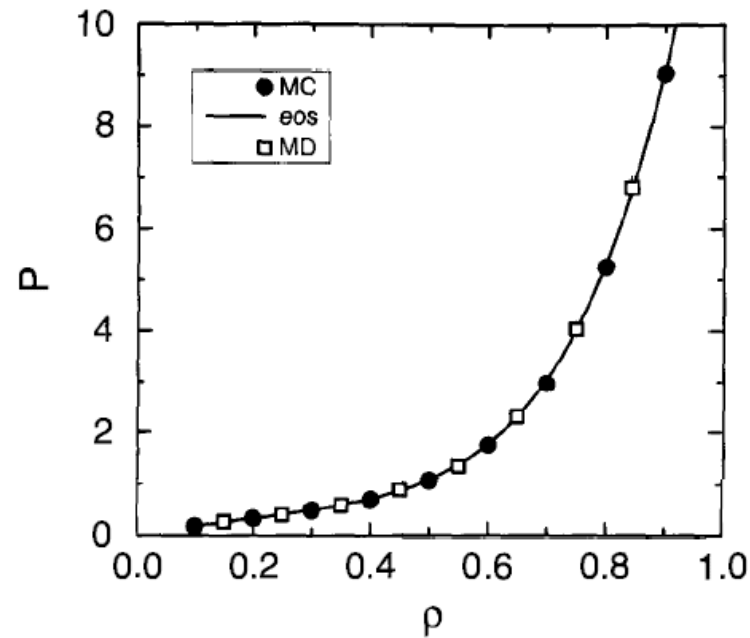
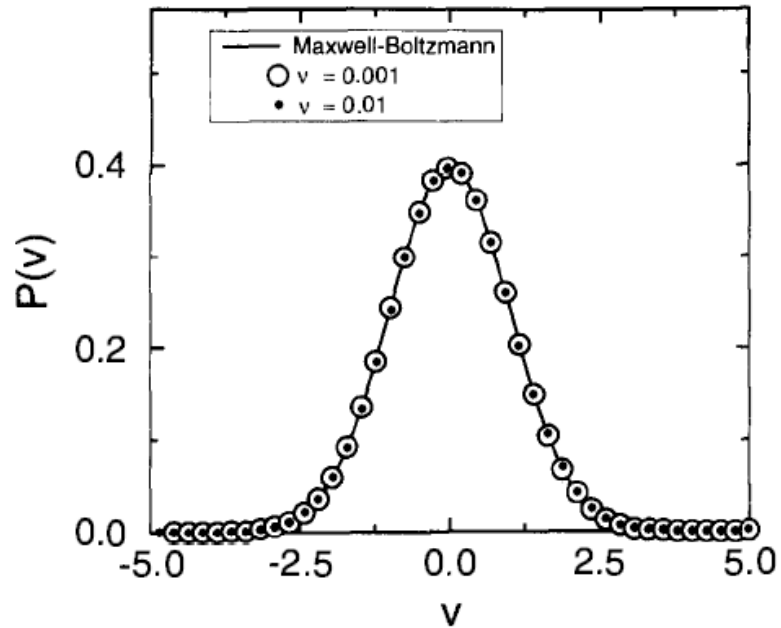
second step velocity Verlet

second update velocity

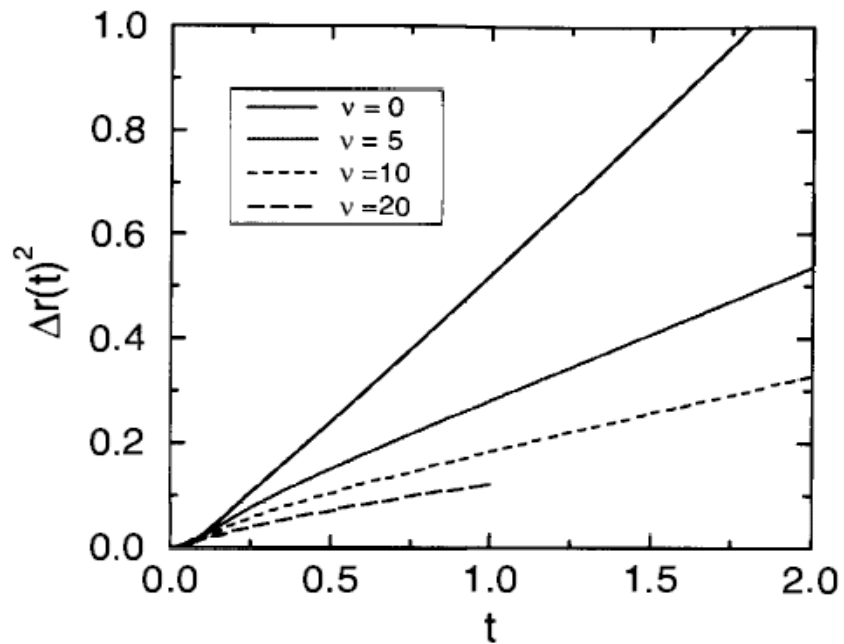
instantaneous temperature  
Andersen heat bath

test for collision with bath  
give particle Gaussian velocity

# Static properties



# Dynamic Properties



# Extended Lagrangian Methods

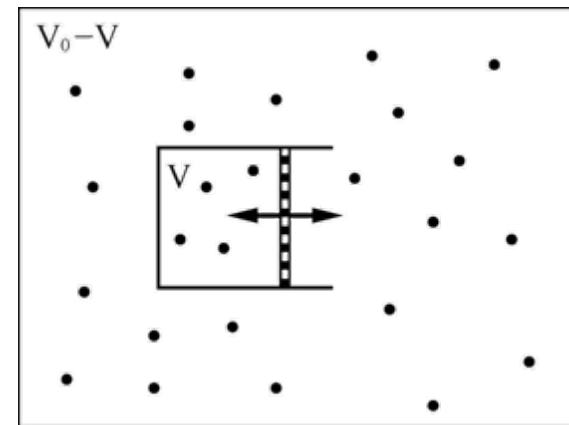
The general idea (we will also discuss this for the constant pressure MD) is to treat an “extended” system, which includes additional degrees of freedom.

In the case that one imposes a constant pressure in this way, one may visualize the additional degree of freedom as the position of the “piston” in the set up we considered Earlier.

An extended Lagrangian is written for the full system.

Lagrangian dynamics:

$$\mathcal{L} = T - V \quad 0 = \frac{d}{dt} \left( \frac{\partial \mathcal{L}}{\partial \dot{q}_j} \right) - \frac{\partial \mathcal{L}}{\partial q_j}$$





# The Nose-Hoover thermostat

The extended Lagrangian, including an additional degree of freedom  $s$  used to obtain the canonical ensemble is:

$$\mathcal{L}_{Nose} = \sum_{i=1}^N \frac{m_i}{2} s^2 \dot{\mathbf{r}}_i^2 - U(\mathbf{r}^N) + \frac{Q}{2} \dot{s}^2 - \frac{L}{\beta} \ln s$$

The conjugate momenta are

$$\mathbf{p}_i = \frac{\partial \mathcal{L}}{\partial \dot{\mathbf{r}}_i} = m_i s^2 \dot{\mathbf{r}}_i \quad p_s = \frac{\partial \mathcal{L}}{\partial \dot{s}} = Q \dot{s}.$$

And the Hamiltonian is  $\mathcal{H}_{Nose} = \sum_{i=1}^N \frac{\mathbf{p}_i^2}{2m_i s^2} + U(\mathbf{r}^N) + \frac{p_s^2}{2Q} + L \frac{\ln s}{\beta}$

We write the (“microcanonical”) partition function for this system as

$$Q_{Nose} = \frac{1}{N!} \int dp_s ds d\mathbf{p}^N d\mathbf{r}^N \delta(E - \mathcal{H}_{Nose})$$

Using scaled variables  $\mathbf{p}' = \mathbf{p}/s$

And defining 
$$\mathcal{H} = \sum_{i=1}^N \frac{\mathbf{p}'_i{}^2}{2m_i} + U(\mathbf{r}^N)$$

And using  $\delta(h(s)) = \delta(s - s_0)/h'(s_0)$

For a function  $h(s)$  with a root at  $s_0$

$$Q_{Nose} = \frac{1}{N!} \int dp_s ds d\mathbf{p}'^N d\mathbf{r}^N \frac{\beta s^{3N+1}}{L} \delta \left\{ s - \exp \left[ -\beta \frac{\mathcal{H}(\mathbf{p}', \mathbf{r}) + \frac{p_s^2}{2Q} - E}{L} \right] \right\}$$

With the choice:  $L = 3N + 1$

$$Q_{Nose} = \frac{C}{N!} \int d\mathbf{p}'^N d\mathbf{r}^N \exp \left[ -\beta \mathcal{H}(\mathbf{p}', \mathbf{r}) \right]$$

With  $r' = r$ ,  $p' = p/s$ ,  $s' = s$ ,  $p'_s = p_s/s$ , and  $t' = t/s$

The equations of motion for the primed (real) variables are:

$$\begin{aligned} \frac{d\mathbf{r}'_i}{dt'} &= \mathbf{p}'_i/m_i \\ \frac{d\mathbf{p}'_i}{dt'} &= -\frac{\partial U}{\partial \mathbf{r}'_i} - (s' p'_s/Q) \mathbf{p}'_i \\ \frac{1}{s} \frac{ds'}{dt'} &= s' p'_s/Q \\ \frac{d(s' p'_s/Q)}{dt'} &= \left( \sum_i p_i'^2/m_i - \frac{3N+1}{\beta} \right) / Q \end{aligned}$$

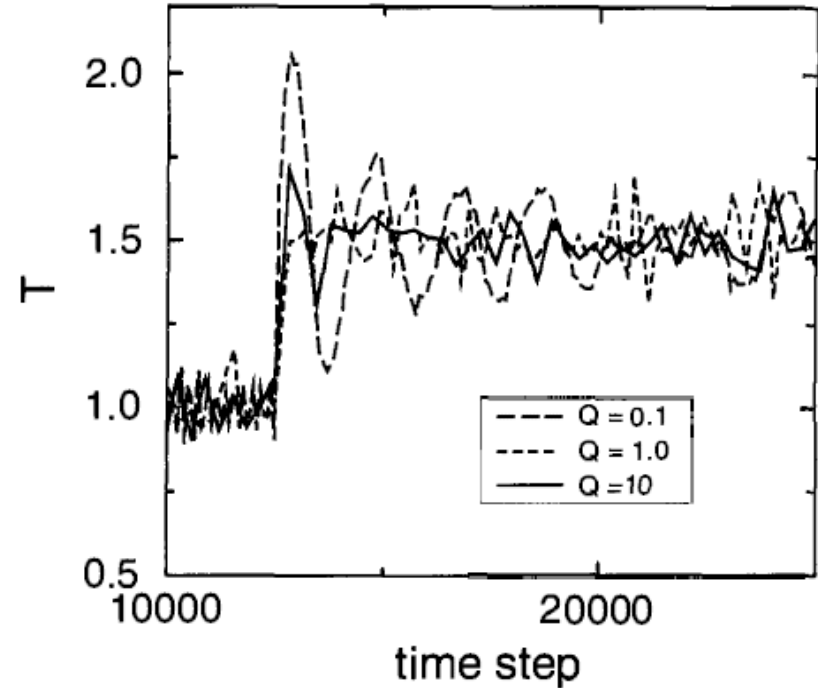
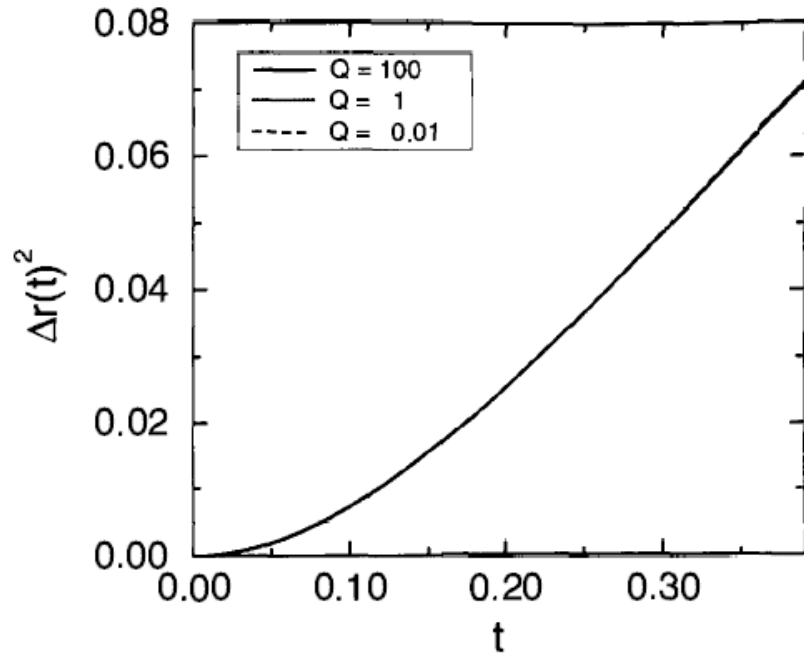
These equations conserve the Hamiltonian

$$\mathcal{H}_{Nose} = \sum_{i=1}^N \frac{p_i'^2}{2m_i} + U(\mathbf{r}'^N) + \frac{s'^2 p_s'^2}{2Q} + L \frac{\ln s'}{\beta}$$

From the relation between the scaled and unscaled variables it is apparent that the variable  $s$  scales the time (in a time dependent way).

$\xi = \dot{s}/s$  is like the friction term in the constraint method, that plays the same role

The Nose-Hoover thermostat generate dynamics that do not critically depend on the thermostat parameter  $Q$ .



In special cases, the Nose-Hoover thermostat fails to generate the canonical distribution, which arises from the conservation of another quantity in addition to  $H_{\text{Nose}}$  (we assumed in deriving the canonical distribution that it was the only conserved quantity). This can be taken care by introducing a series of thermostat variables that are coupled to each other (Nose-Hoover chains).

# Constant pressure molecular dynamics

Along the same lines as constant pressure MD, in this case we write an extended Lagrangian, with two additional variables Instead of one. The equation of motion is given by:

$$\dot{\mathbf{r}}_i = \frac{\mathbf{p}_i}{m_i} + \frac{p_\epsilon}{W} \mathbf{r}_i$$
$$\dot{\mathbf{p}}_i = \mathbf{F}_i - \left(1 + \frac{1}{N}\right) \frac{p_\epsilon}{W} \mathbf{p}_i - \frac{p_\xi}{Q} \mathbf{p}_i$$

The variable that determines volume change is  $\epsilon = \ln(V/V_0)$   
And the corresponding equations of motion are

$$\dot{V} = \frac{dV p_\epsilon}{W},$$

$$\dot{p}_\epsilon = dV (P_{int} - P_{ext}) + \frac{1}{N} \sum_{i=1}^N \frac{\mathbf{p}_i^2}{m_i} - \frac{p_\xi}{Q} p_\epsilon,$$

$P_{int}$  and  $P_{ext}$  are the internal pressure (from virial) and the applied pressure

# Some Special Cases and Techniques

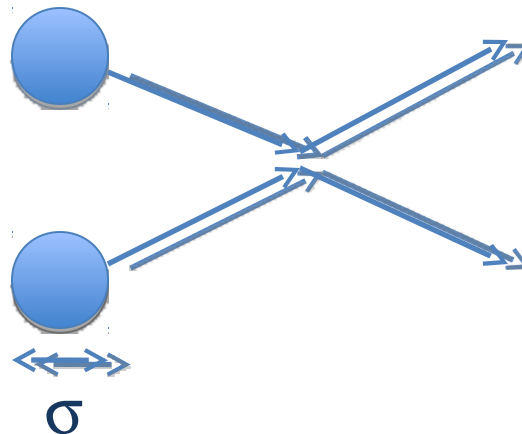
There are special cases where specific methods that are distinct from those discussed above are used. Some of these are:

1. Discontinuous potentials (hard spheres etc): Event driven molecular dynamics
2. Systems with long range interactions: Ewald Sum
3. Molecular systems: SHAKE

# Event Driven Molecular Dynamics

In systems interacting with hard core, or step potentials, the momentum change is discontinuous upon contact, and therefore the fixed time step integration methods are not useful.

Instead, the system is propagated from one collision event (hard sphere case) to the next, and the momentum change implemented at the collision, with the following steps:





1. Locate next collision. For a pair  $i, j$ , the condition for collision time is

$$|\mathbf{r}_{ij} + \mathbf{v}_{ij}t_{ij}| = \sigma$$

$$v_{ij}^2 t_{ij}^2 + 2b_{ij}t_{ij} + r_{ij}^2 - \sigma^2 = 0 \quad b_{ij} = \mathbf{r}_{ij} \cdot \mathbf{v}_{ij}$$

2. Cases:

i. If  $b_{ij} > 0$ , particles are moving away.

ii. Root of equation are complex if  $b_{ij}^2 - v_{ij}^2(r_{ij}^2 - \sigma^2) < 0$

iii. Else smaller of the roots is the collision time:

$$t_{ij} = \frac{-b_{ij} - (b_{ij}^2 - v_{ij}^2(r_{ij}^2 - \sigma^2))^{1/2}}{v_{ij}^2}$$

3. Find smallest collision time among all pairs.
4. Propagate all particles up to the collision time.
5. For colliding particles, invert velocities in the direction of Collision:

$$\mathbf{v}_i = \mathbf{v}_i^o + \delta\mathbf{v} \quad \mathbf{v}_j = \mathbf{v}_j^o - \delta\mathbf{v} \quad \delta\mathbf{v} = -\frac{b_{ij}}{\sigma^2}\mathbf{r}_{ij}$$

6. Repeat. Keep track of time elapsed.

Since the dynamics is updated according to the collision events, this method is termed event driven molecular dynamics.

# Molecular Systems

For molecular systems, we have both intermolecular and intramolecular distances and forces to contend with.

Intramolecular vibrations are much faster than intermolecular motion, and often not of interest.

Hence, use rigid molecular models. Eg, SPC/E model of water:

OH distance :  $1 \text{ \AA}$

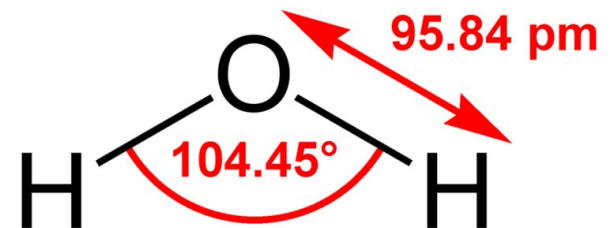
HOH angle :  $109.47^\circ$

$q_H = 0.4238 e$

$q_O = -2q_H$

O-O interaction via LJ

$\sigma = 3.166 \text{ \AA}$   $\epsilon = 0.6502 \frac{\text{KJ}}{\text{mol}}$



OH distances and HOH angle do not change during simulation.

How to handle such fixed distances and angles?

1. Work with centre of mass + rotational degrees of freedom.  
Too complicated.

2. Add constraint forces to keep shape of molecule fixed.

$$F_{i\alpha} = F_{i\alpha}^{free} + F_{i\alpha}^{constraint}$$

Define constraint parameter  $\chi_{12} = r_{12}^2 - d_{12}^2 = 0$

The force, including constraints is

$$F_1 = F_1^{free} + \frac{1}{2}\lambda_{12} \nabla_{r_1}\chi_{12}$$

## Implementing all the constraints

$$F_1^c = \lambda_{12}\vec{r}_{12} - \lambda_{31}\vec{r}_{31}$$

$$F_2^c = \lambda_{23}\vec{r}_{23} - \lambda_{12}\vec{r}_{12}$$

$$F_3^c = \lambda_{31}\vec{r}_{31} - \lambda_{23}\vec{r}_{23}$$

The Verlet update will be

$$r_\alpha(t + \delta t) = r'_\alpha(t + \delta t) + \frac{\delta t^2 F_\alpha^c}{m_\alpha}$$

$$r_{12}(t + \delta t) = r'_{12}(t + \delta t) + \delta t^2(m_1^{-1} + m_2^{-1})\lambda_{12}r_{12}(t) - \delta t^2 m_2^{-1}\lambda_{23}(t) - \delta t^2 m_1^{-1}\lambda_{31}r_{31}(t)$$

where primed coordinates are updates without constraint.

The constraints, written as

$$|r_{12}(t + \delta t)|^2 = |r_{12}(t)|^2 = d_{12}^2$$

yield quadratic equations for the parameters  $\lambda$ . But instead, we linearize and iteratively solve the equation to convergence.

$$r_{i\alpha}(t + \delta t) = r_{i\alpha}(t + \delta t) + \frac{1}{2} \left( \frac{m_\beta}{m_\alpha + m_\beta} \right) \left[ \frac{r_{\alpha\beta}^2(t + \delta t) - d_{\alpha\beta}^2}{r_{\alpha\beta}(t + \delta t)} \right]$$

$$r_{i\beta}(t + \delta t) = r_{i\beta}(t + \delta t) - \frac{1}{2} \left( \frac{m_\alpha}{m_\alpha + m_\beta} \right) \left[ \frac{r_{\alpha\beta}^2(t + \delta t) - d_{\alpha\beta}^2}{r_{\alpha\beta}(t + \delta t)} \right]$$

# Long Range Interactions: Ewald Sum

Normally, we truncate the potential so that we need to consider only the minimum image.

This method will not work for long range interactions (Coulomb interactions)

In the Ewald sum method, the potential due to other charges, and all their periodic images, is carried out efficiently.

This is done by introducing a “screening” charge distribution around each point charge, and adding a second compensating charge distribution. The Coulomb interactions with screened charges are now short ranged, and can be treated as such.

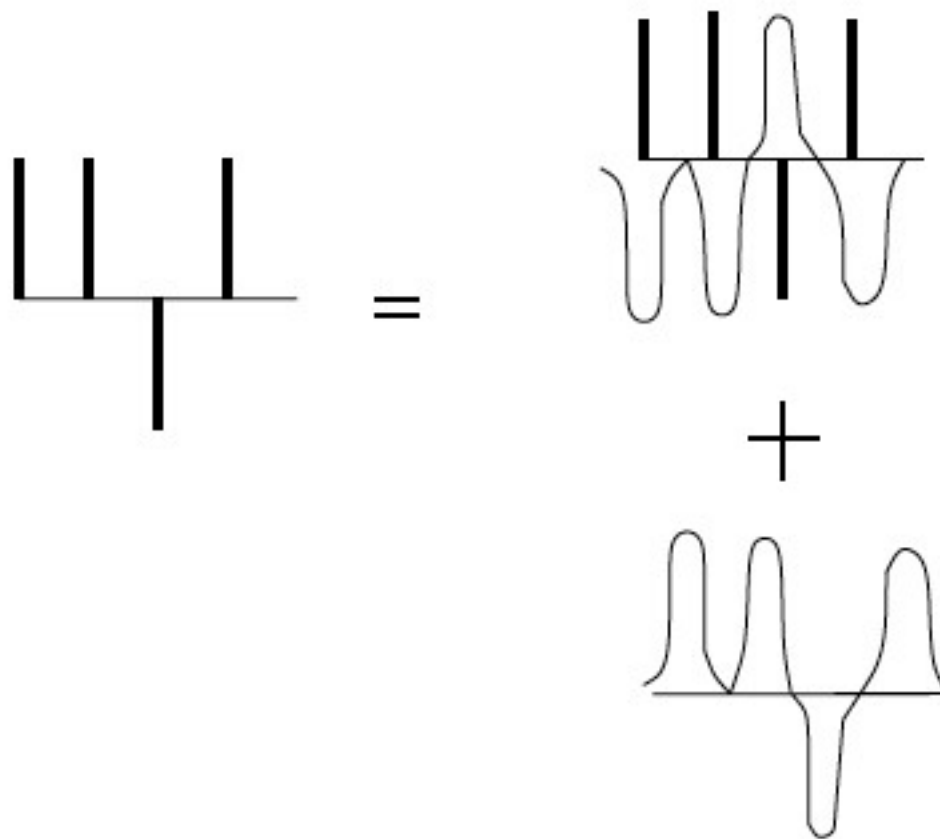
The potential due to the compensating charges (and all its images) is efficiently calculated in Fourier space, IF we include also the charge distributions in the primary simulation cell.

This introduces a “self” interaction term, that needs to be compensated.





Schematic:



## Interaction with compensating charges

First we calculate the potential at  $r_i$  due to periodic array of compensating charges

$$\rho_{comp}(r) = \sum_{\mathbf{n}, j} q_j \left(\frac{\alpha}{\pi}\right)^{3/2} \exp[-\alpha |\mathbf{r} - (\mathbf{r}_j + \mathbf{nL})|^2]$$

The Fourier transform of the charge distribution is:

$$\begin{aligned} \rho_{comp}(\mathbf{k}) &= \frac{1}{V} \int_{all\ space} d\mathbf{r} \exp[-i\mathbf{k}\cdot\mathbf{r}] \sum_{j=1}^N q_j \left(\frac{\alpha}{\pi}\right)^{3/2} \exp[-\alpha |\mathbf{r} - \mathbf{r}_j|^2] \\ &= \frac{1}{V} \sum_{j=1}^N \exp(-i\mathbf{k}\cdot\mathbf{r}_j) \exp\left(-\frac{k^2}{4\alpha}\right) \end{aligned}$$

We obtain the electrostatic potential by solving the Poisson equation in Fourier space

$$\begin{aligned} -\nabla^2 \phi(r) &= 4\pi \rho(r) & \phi(k) &= \frac{4\pi}{k^2} \rho(k) \\ k^2 \phi(k) &= 4\pi \rho(k) \end{aligned}$$

The potential from the compensating charges then is:

$$\phi_{comp}(k) = \frac{4\pi}{k^2} \frac{1}{V} \sum_{j=1}^N q_j \exp(-i\mathbf{k} \cdot \mathbf{r}_j) \exp\left(-\frac{k^2}{4\alpha}\right)$$

In real space

$$\phi_{comp}(\mathbf{r}) = \sum_{k \neq 0} \sum_{j=1}^N \frac{4\pi q_j}{k^2} \exp[i\mathbf{k} \cdot (\mathbf{r} - \mathbf{r}_j)] \exp\left(-\frac{k^2}{4\alpha}\right)$$

The potential for all the charges is

$$\begin{aligned} U_{comp} &= \frac{1}{2} \sum_i q_i \phi_1(r_i) \\ &= \frac{1}{2V} \sum_{k \neq 0} \sum_{i,j=1}^N \frac{4\pi q_j q_i}{k^2} \exp[i\mathbf{k} \cdot (\mathbf{r}_i - \mathbf{r}_j)] \exp\left(-\frac{k^2}{4\alpha}\right) \\ &= \frac{V}{2} \sum_{k \neq 0} \frac{4\pi}{k^2} |\rho(k)|^2 \exp\left(-\frac{k^2}{4\alpha}\right). \end{aligned}$$

with

$$\rho(k) = \frac{1}{V} \sum_{i=1}^N q_i \exp[i\mathbf{k} \cdot \mathbf{r}_i]$$

## Correction for self interaction

The previous summation includes also the compensating charge distribution at the location of the charge. To subtract it, we calculate the potential due to a gaussian charge distribution at a distance  $r$ :

$$\phi_{Gauss}(r) = q_i \frac{\text{erf}(\sqrt{\alpha}r)}{r}$$

The self interaction therefore is

$$\begin{aligned} U_{self} &= \frac{1}{2} \sum_{i=1}^N q_i \phi_{self}(r_i) \\ &= \left(\frac{\alpha}{\pi}\right)^{1/2} \sum_{i=1}^N q_i^2 \end{aligned}$$

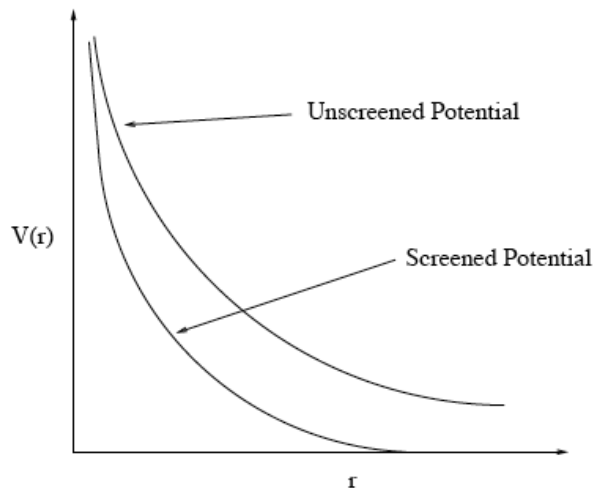
## Interaction with screened charges

The potential due to screened charges is

$$\begin{aligned}\phi_{\text{screened}}(r) &= \frac{q_i}{r} - \frac{q_i}{r} \operatorname{erf}(\sqrt{\alpha}r) \\ &= \frac{q_i}{r} \operatorname{erfc}(\sqrt{\alpha}r)\end{aligned}$$

The energy

$$U_{\text{screened}} = \frac{1}{2} \sum_{i,j} q_i q_j \operatorname{erfc}(\sqrt{\alpha}r_{ij}) / r_{ij}$$



## The Ewald sum energy

Combining all these terms

$$\begin{aligned} U_{Coul} &= \frac{V}{2} \sum_{k \neq 0} \frac{4\pi}{k^2} |\rho(k)|^2 \exp\left(-\frac{k^2}{4\alpha}\right) \\ &- \left(\frac{\alpha}{\pi}\right)^{1/2} \sum_{i=1}^N q_i^2 \\ &+ \frac{1}{2} \sum_{i \neq j} q_i q_j \operatorname{erfc}(\sqrt{\alpha}r)/r_{ij} \end{aligned}$$