

MD problems we will deal with

① real interactions are more complicated.

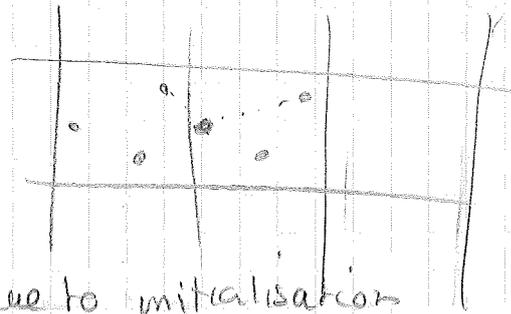
Quantum Mechanics; electron clouds of atoms interact too much work if you want to do many particles

⇒ realistic effective classical force fields / potentials.
- combination: Quantum MD (still pretty expensive)

⇒ unlike in previous examples usually smooth interaction

② $N \leq N_A$, $t \leq 1s$ $d \ll 1m$

periodic boundary conditions (nearest image convention)
be careful if correlation length $> d$



finite-size effects

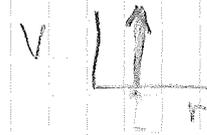
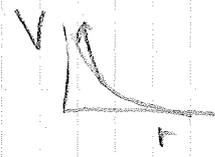
averaging over finite time ⇒ errors due to initialisation & statistical errors

③ Integration algorithms have finite accuracy

Example: smooth interactions in a gas

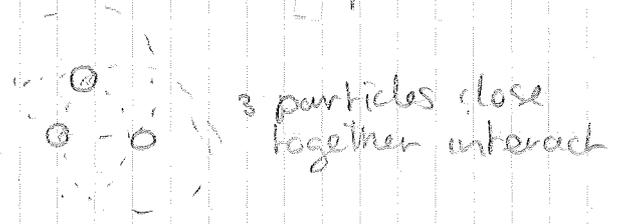
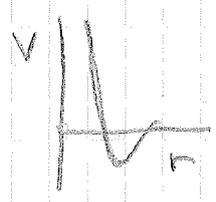
hard spheres & instantaneous only (no 3-particle interactions) ↔ low density.

more realistic soft repulsion



attractive interaction

(f.e. van der Waals)

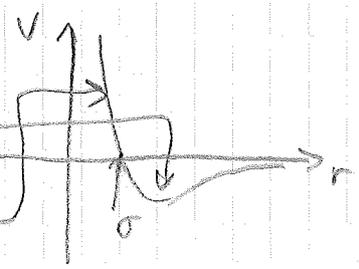


Your first simulation project: Argon liquid
dense, and can actually solidify

Lennard-Jones $V(\vec{r}_1, \vec{r}_2) = U_{LJ}(|\vec{r}_1 - \vec{r}_2|)$

$$U_{LJ}(r) = 4\epsilon \left[\left(\frac{\sigma}{r}\right)^{12} - \left(\frac{\sigma}{r}\right)^6 \right]$$

↑ strong repulsive part
↑ weak attractive part



(LJ continued)

$$U_L = 0 \text{ at } \sigma$$

Minimum at

$$\frac{\partial U_L}{\partial r} = 0 = 4\epsilon \left[12 \left(\frac{\sigma}{r}\right)^{11} - 6 \left(\frac{\sigma}{r}\right)^5 \right] \frac{\partial}{\partial r} \left(\frac{\sigma}{r}\right)$$

$$\Rightarrow 2 \left(\frac{\sigma}{r}\right)^6 = 1 \Rightarrow r_{\min} = \sigma^{2/6}$$

$$U_L(r_{\min}) = -\epsilon$$

LJ very common; good for argon too

$$\epsilon = 1.654 \cdot 10^{-21} \text{ J} \quad \sigma = 3.405 \text{ \AA}$$

+ minimum image convention + integration = your simulation

Note on efficiency when you calculate forces

$$F_L = -\frac{\partial U}{\partial r} = 4\epsilon \left[12 \frac{\sigma^{11}}{r^{11}} - 6 \frac{\sigma^5}{r^5} \right] \sigma \frac{r}{r^3}$$

$$= 4\epsilon \left[12 \frac{\sigma^{12}}{r^{14}} - 6 \frac{\sigma^6}{r^8} \right] \vec{r}$$

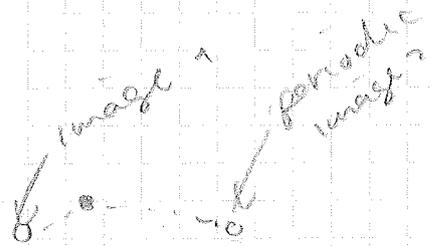
↑ ↑ even powers, no expensive $\text{sqrt}((r_i - r_j)^2)$

Note: You can now get in trouble with periodic boundaries, and interact with several images of a particle.

sometimes that's ok, but not here

Nearest image convention:

interact only with nearest image.



Argon liquid model: 27, N atoms, periodic boundaries

variables $(\vec{r}_1, \vec{r}_2, \vec{r}_3, \dots, \vec{r}_N, \vec{v}_1, \vec{v}_2, \dots, \vec{v}_N) = \gamma$ ← dynamical system

eqs of motion $\dot{\vec{r}}_i = \vec{v}_i$
 $\dot{\vec{v}}_i = - \sum_j \frac{\partial V(\vec{r})}{m_i \partial \vec{r}_i} \Big|_{\vec{r} = \vec{r}_i - \vec{r}_j} = \vec{a}_i$ \downarrow
 $\dot{\gamma} = f(\gamma, t)$

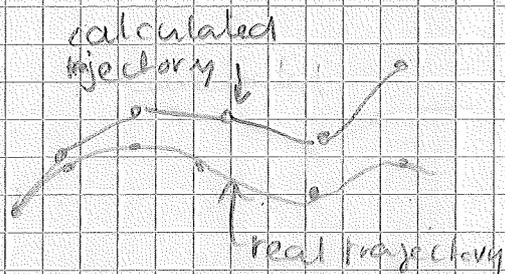
How to solve? Small time steps h

Naive approach (Euler)

$$\gamma(t+h) = \gamma(t) + h f(\gamma, t) + O(h^2)$$

$$\left\{ \begin{aligned} r(t+h) &= r(t) + h v(t) \quad (+ O(h^2)) \\ v(t+h) &= v(t) + h a(r(t)) \quad (+ O(h^2)) \end{aligned} \right.$$

$$\left\{ \begin{aligned} r(t+h) &= r(t) + h v(t) \quad (+ O(h^2)) \\ v(t+h) &= v(t) + h a(r(t)) \quad (+ O(h^2)) \end{aligned} \right.$$



$$\begin{aligned} E(t+h) &= \frac{1}{2} m [v(t) + h a(r(t))]^2 + V(r(t) + h v(t)) \\ &= \frac{1}{2} m v(t)^2 + m h a(r(t)) \cdot v(t) + \frac{1}{2} m h^2 a(r(t))^2 \\ &\quad + V(r(t)) + h v(t) V'(r(t)) + \frac{1}{2} (h v(t))^2 V''(r(t)) \\ &= E(t) + O(h^2) \end{aligned}$$

also error in γ is ht

This is terrible and very unstable

DO NOT USE

We can do better than Euler; Verlet
in terms of $v(t-h), a(t-h)$
in terms of $v(t), a(t-h)$



backwards / inverse (forwards)

$$r(t+h) = r(t) + h v(t) + \frac{1}{2} h^2 a(t) + \frac{1}{6} h^3 \ddot{a}(t) + O(h^4)$$

$$+ r(t-h) = r(t) - h v(t) + \frac{1}{2} h^2 a(t) - \frac{1}{6} h^3 \ddot{a}(t) + O(h^4)$$

(addition)

$$r(t+h) = -r(t-h) + 2r(t) + h^2 a(t) + O(h^4) \quad (1)$$

(subtraction)

$$v(t) = \frac{1}{2h} [r(t+h) - r(t-h)] + O(h^2) \quad (2)$$

↑ from $\frac{1}{6} h^3 \ddot{a}(t) \frac{1}{2h}$

velocity is byproduct, not in expression for $r(t+h)$

What happens to errors after long time?

use induction for error in r ($n = \frac{t}{h}$)

$$\text{err}(0) = 0$$

$$\text{err}(1) = O(h^4)$$

$$\text{err}(n+1) = -\text{err}(n-1) + 2\text{err}(n) + h^2 \text{err}(n) + O(h^4)$$

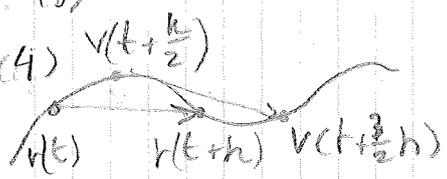
Solution: $\text{err}(n) \propto r(n+1)$

$$\Rightarrow \text{error after time } t = \left(\frac{t}{h}\right)^2 O(h^4) \approx t^2 h^2$$

Other Verlet incarnations

Leap-frog $v(t+h/2) = v(t-h/2) + h a(t)$ (3)

$r(t+h) = r(t) + h v(t+h/2)$ (4)



(4) is similar to (2) with $h \rightarrow 2h$

(3) is reformulation of (1)

Velocity verlet

$r(t+h) = r(t) + h v(t) + h^2 a(t)/2$

$v(t+h) = v(t) + h [a(t+h) + a(t)]/2$

Implement as
 $\tilde{v}(t) = v(t) + h a(t)/2$
 $r(t+h) = r(t) + h \tilde{v}(t)$
 $v(t+h) = \tilde{v}(t) + h a(t+h)$

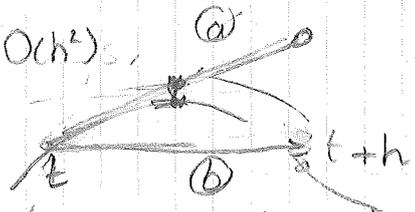
use this for programming average of forward & backward

This was for simple eqs of motion with Hamiltonian dynamics.

Can include in Verlet $\dot{v}(t) = a(r(t)) - \gamma v(t)$ viscous friction (common addition in models, for instance Langevin thermostat)

Problem: other non-Hamiltonian terms not always easy. $\dot{y} = f(y, t)$

(a) $y(t+h) = y(t) + h f(y(t))$ error $O(h^2)$



for time dependency, see appendix

(b) $y(t+h) = y(t) + \frac{1}{2} h f(y(t))$

$y(t+h) = y(t) + h f(y(t + \frac{1}{2}h))$ midpoint rule
 Taylor expansion: $= y(t) + k_2 + O(h^3)$, $k_2 = h f(y(t) + \frac{1}{2}k_1)$

$y(t+h) = y(t) + h f(y(t) + \frac{1}{2} h f(y(t)))$
 $= y(t) + h [f(y(t)) + \frac{df}{dy} \frac{1}{2} h f(y(t))]$
 $= y(t) + h \dot{y}(t) + \frac{1}{2} h^2 \ddot{y}(t)$

error is $O(h^3)$

(c) Even more midpoints:

$k_1 = h f(y(t))$ $k_2 = h f(y(t) + \frac{1}{2}k_1)$
 $k_3 = h f(y(t) + \frac{1}{2}k_2)$ $k_4 = h f(y(t) + k_3)$
 $y(t+h) = y(t) + \frac{1}{6}(k_1 + k_2 + k_3 + k_4) + O(h^5)$

4th order Runge-Kutta.
 (you can take this further)

big advantage!
 efficient with variable time step (so that you can save computation time in intermittent systems like dilute gasses)
 RK-Fehlberg
 compare RK4, RK5

very general method more work than Verlet end lecture 2.