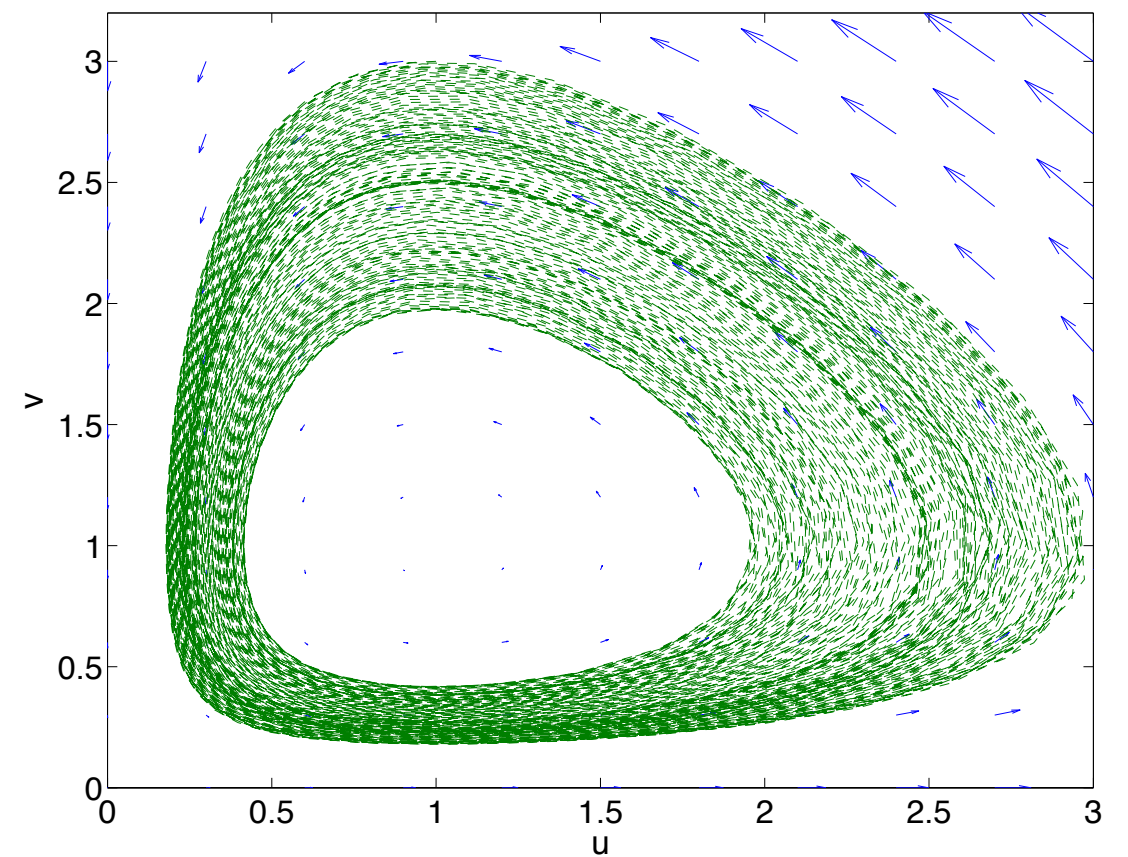


Splitting methods and rigid body integration

Ralf Banisch



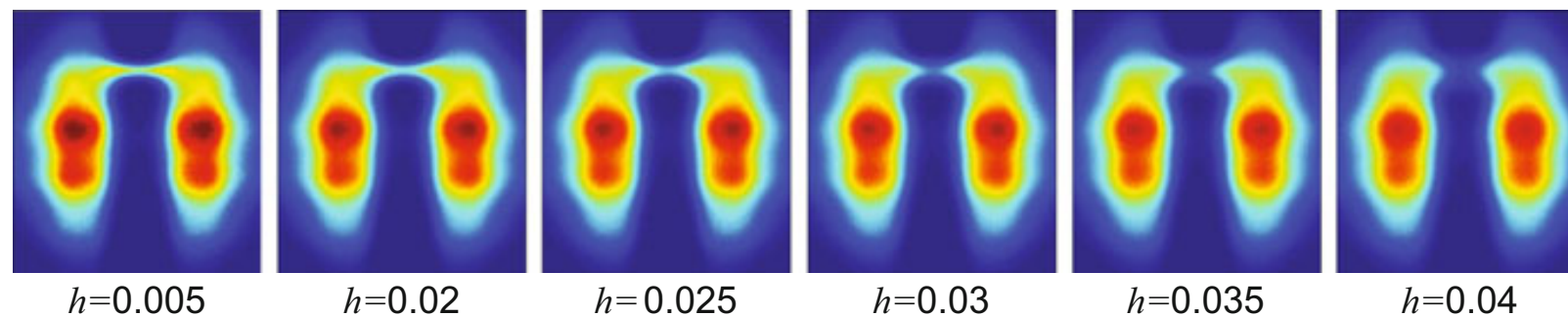
Outline

- Langevin Dynamics
- Numerical integration, splitting strategies
- The Mercedes Benz potential as a model for water like particles
- Rigid bodies
- The DLM algorithm

Book 'Molecular Dynamics' by
Ben Leimkuhler & Charles Matthews

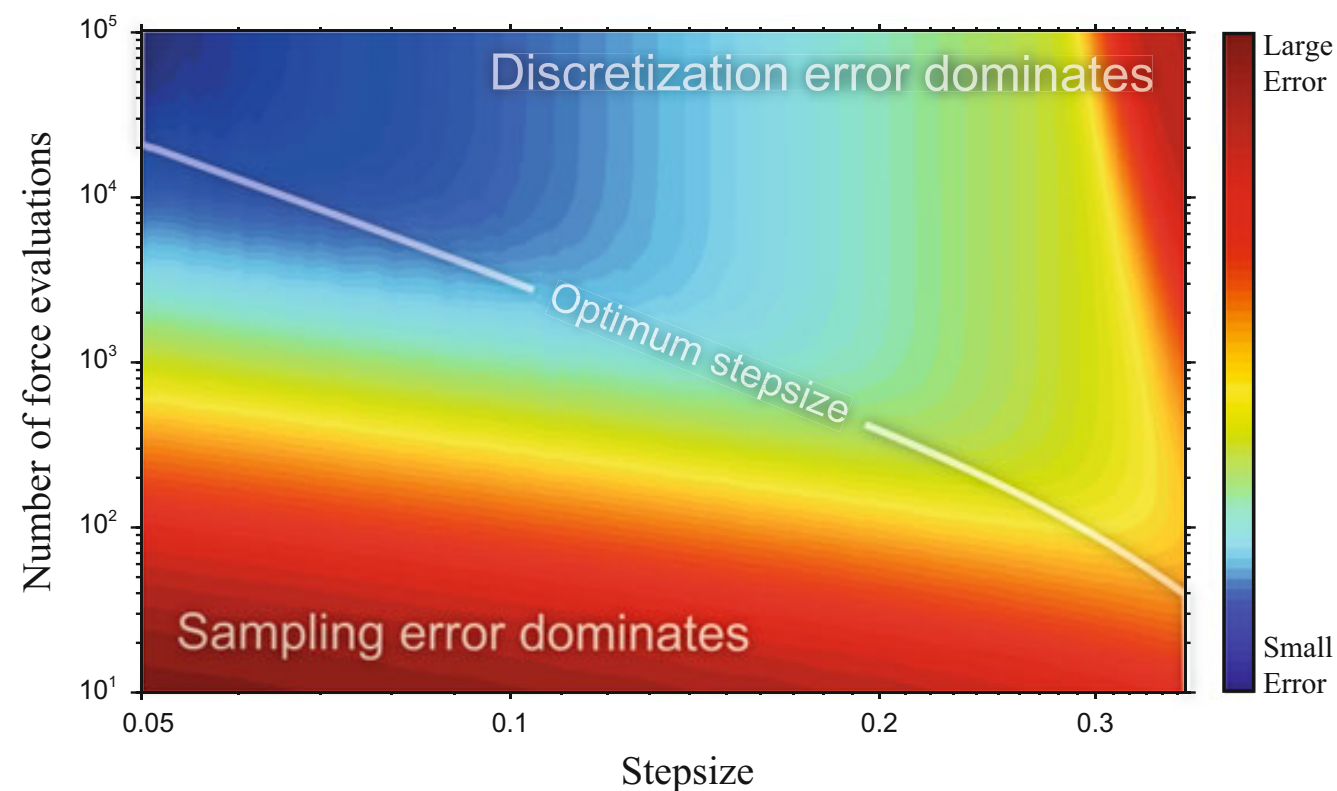
Numerical integration in MD: Sampling error vs discretisation error

- Discretisation error due to introduction of discretisation time step h :



Picture from
[Leimkuhler, Matthews '15]

- Sampling error vs. discretisation error:



Picture from
[Leimkuhler, Matthews '15]

Langevin dynamics

- Langevin dynamics

$$d\mathbf{q} = \mathbf{M}^{-1}\mathbf{p}dt,$$

$$d\mathbf{p} = -\nabla U(\mathbf{q})dt - \gamma\mathbf{p}dt + \sqrt{2\gamma k_B T}\mathbf{M}^{1/2}d\mathbf{W}.$$

- Langevin dynamics is ergodic:

$$(q, p) \sim Z^{-1}e^{-H(q,p)}$$

- Splitting strategy:

$$d\begin{bmatrix} \mathbf{q} \\ \mathbf{p} \end{bmatrix} = \underbrace{\begin{bmatrix} \mathbf{M}^{-1}\mathbf{p} \\ \mathbf{0} \end{bmatrix}}_A dt + \underbrace{\begin{bmatrix} \mathbf{0} \\ -\nabla U(\mathbf{q}) \end{bmatrix}}_B dt + \underbrace{\begin{bmatrix} \mathbf{0} \\ -\gamma\mathbf{p}dt + \sigma\mathbf{M}^{1/2}d\mathbf{W} \end{bmatrix}}_O,$$

Splitting strategies

- Splitting strategy:

$$d \begin{bmatrix} \mathbf{q} \\ \mathbf{p} \end{bmatrix} = \underbrace{\begin{bmatrix} \mathbf{M}^{-1}\mathbf{p} \\ \mathbf{0} \end{bmatrix} dt}_{\text{A}} + \underbrace{\begin{bmatrix} \mathbf{0} \\ -\nabla U(\mathbf{q}) \end{bmatrix} dt}_{\text{B}} + \underbrace{\begin{bmatrix} \mathbf{0} \\ -\gamma \mathbf{p} dt + \sigma \mathbf{M}^{1/2} d\mathbf{W} \end{bmatrix}}_{\text{O}},$$

- Update maps for the individual pieces (O part is solved exactly):

$$\mathcal{U}_h^{\text{A}}(\mathbf{q}, \mathbf{p}) = (\mathbf{q} + h\mathbf{M}^{-1}\mathbf{p}, \mathbf{p}),$$

$$\mathcal{U}_h^{\text{B}}(\mathbf{q}, \mathbf{p}) = (\mathbf{q}, \mathbf{p} - h\nabla U(\mathbf{q})),$$

$$\mathcal{U}_h^{\text{O}}(\mathbf{q}, \mathbf{p}) = (\mathbf{q}, e^{-\gamma h}\mathbf{p} + \sqrt{k_{\text{B}}T(1 - e^{-2\gamma h})}\mathbf{M}^{1/2}\mathbf{R}),$$

- We can define a family of schemes codified as strings with the alphabet “ABO”, e.g.

$$\mathcal{U}_h^{\text{[ABO]}} = \mathcal{U}_h^{\text{O}} \circ \mathcal{U}_h^{\text{B}} \circ \mathcal{U}_h^{\text{A}}.$$

$$\mathcal{U}_h^{\text{[BABO]}} = \mathcal{U}_h^{\text{O}} \circ \mathcal{U}_{h/2}^{\text{B}} \circ \mathcal{U}_h^{\text{A}} \circ \mathcal{U}_{h/2}^{\text{B}}.$$

Splitting strategies

- Symmetric schemes like “ABOBA” typically have more favourable properties compared to non-symmetric ones, e.g. “ABAO”
- Some schemes and explicit maps:

[[BAOAB]]	[[OBABO]]	[[ABOBA]]
$\mathbf{p}_{n+1/2} = \mathbf{p}_n - (h/2)\nabla U(\mathbf{q}_n),$	$\mathbf{p}_{n+1/2} = e^{-h\gamma/2}\mathbf{p}_n + \zeta_1\mathbf{M}^{1/2}\mathbf{R}_n,$	$\mathbf{q}_{n+1/2} = \mathbf{q}_n + (h/2)\mathbf{M}^{-1}\mathbf{p}_n,$
$\mathbf{q}_{n+1/2} = \mathbf{q}_n + (h/2)\mathbf{M}^{-1}\mathbf{p}_{n+1/2},$	$\hat{\mathbf{p}}_{n+1/2} = \mathbf{p}_{n+1/2} - (h/2)\nabla U(\mathbf{q}_n),$	$\mathbf{p}_{n+1/2} = \mathbf{p}_n - (h/2)\nabla U(\mathbf{q}_{n+1/2}),$
$\hat{\mathbf{p}}_{n+1/2} = e^{-h\gamma}\mathbf{p}_{n+1/2} + \zeta_2\mathbf{M}^{1/2}\mathbf{R}_n,$	$\mathbf{q}_{n+1} = \mathbf{q}_n + h\mathbf{M}^{-1}\hat{\mathbf{p}}_{n+1/2},$	$\hat{\mathbf{p}}_{n+1/2} = e^{-h\gamma}\mathbf{p}_{n+1/2} + \zeta_2\mathbf{M}^{1/2}\mathbf{R}_n,$
$\mathbf{q}_{n+1} = \mathbf{q}_{n+1/2} + (h/2)\mathbf{M}^{-1}\hat{\mathbf{p}}_{n+1/2},$	$\hat{\mathbf{p}}_{n+1} = \hat{\mathbf{p}}_{n+1/2} - (h/2)\nabla U(\mathbf{q}_{n+1}),$	$\mathbf{p}_{n+1} = \hat{\mathbf{p}}_{n+1/2} - (h/2)\nabla U(\mathbf{q}_{n+1/2}),$
$\mathbf{p}_{n+1} = \hat{\mathbf{p}}_{n+1/2} - (h/2)\nabla U(\mathbf{q}_{n+1}),$	$\mathbf{p}_{n+1} = e^{-h\gamma/2}\hat{\mathbf{p}}_{n+1} + \zeta_1\mathbf{M}^{1/2}\mathbf{R}_{n+1/2},$	$\mathbf{q}_{n+1} = \mathbf{q}_{n+1/2} + (h/2)\mathbf{M}^{-1}\mathbf{p}_{n+1},$

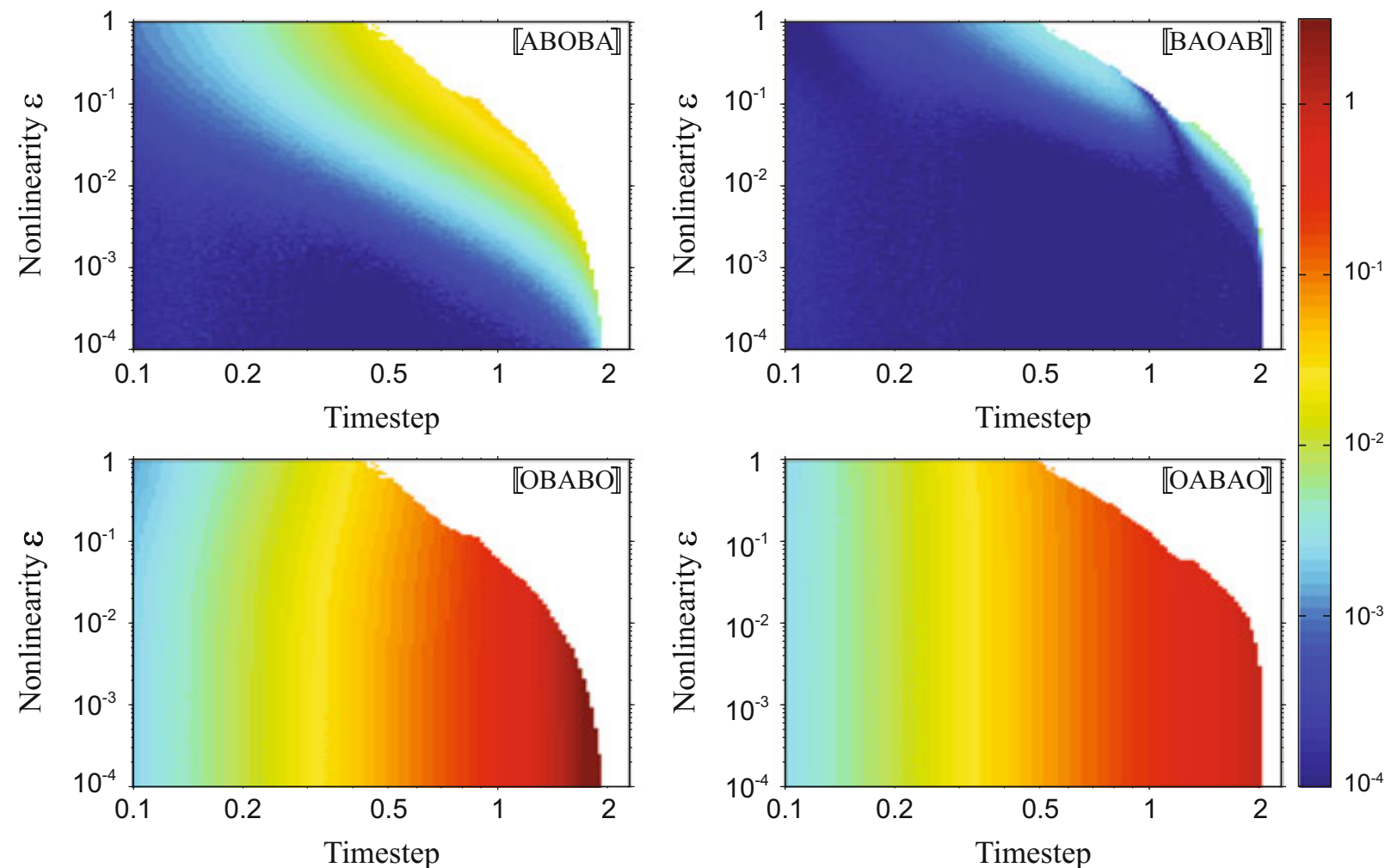
$$\zeta_j = [\mathbf{k}_B T (1 - e^{-j\gamma h})]^{1/2}, \quad \mathbf{R}_n, \mathbf{R}_{n+1/2} \text{ are } N(0,1) \text{ random variables}$$

- One derives error bounds by considering perturbative expansions of e.g.

$$e^{h\hat{\mathcal{L}}^\dagger} = e^{\frac{h}{2}\mathcal{L}_X^\dagger} e^{\frac{h}{2}\mathcal{L}_Y^\dagger} e^{h\mathcal{L}_Z^\dagger} e^{\frac{h}{2}\mathcal{L}_Y^\dagger} e^{\frac{h}{2}\mathcal{L}_X^\dagger},$$

Splitting strategies

- Different schemes can have very different discretisation error!



Picture from
[Leimkuhler, Matthews '15]

Fig. 7.3 Each pixel in the above grids marks the result of a numerical experiment using one of the four second order (in stepsize) symmetric Langevin discretization schemes. A pixel's color denotes the absolute error of $\langle q^2 \rangle_h$ from an experiment conducted at the given stepsize h (horizontal) and nonlinearity parameter ϵ (vertical) defining the potential energy in (7.13), with white pixels indicating instability. Experiments were taken over a sufficiently long time interval such that the observed errors are the result of discretization error, rather than sampling error

Ergodicity

Langevin dynamics (friction parameter $\gamma > 0$)

$$\begin{cases} dq_t = M^{-1} p_t dt, \\ dp_t = -\nabla V(q_t) dt - \gamma M^{-1} p_t dt + \sqrt{\frac{2\gamma}{\beta}} dW_t. \end{cases}$$

- Unique invariant measure μ
- Convergence of ergodic averages over one trajectory

$$\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t \varphi(q_s, p_s) ds = \mathbb{E}_\mu(\varphi) = \int_{\mathcal{E}} \varphi(q, p) d\mu(q, p), \text{ a.s.}$$

- Rate of convergence \rightarrow Central limit theorem : define $\Pi\varphi := \varphi - \mathbb{E}_\mu(\varphi)$

$$\sqrt{t} \left(\frac{1}{t} \int_0^t \Pi\varphi(q_t, p_t) dt \right) \xrightarrow[t \rightarrow +\infty]{\text{law}} \mathcal{N}(0, \sigma^2)$$

with asymptotic variance $\sigma^2(\varphi) = 2 \int_0^{+\infty} \mathbb{E}(\Pi\varphi(q_t, p_t) \Pi\varphi(q_0, p_0)) dt$

Ergodicity

- Need to check for ergodicity by computing trajectory averages of observables whose exact ergodic averages are known.

Kinetic temperature

$$T = \text{Av}_\beta \left(\frac{p^2}{mk_B} \right).$$

configurational temperature

$$\frac{\text{Av}_\beta(\|\nabla U\|^2)}{\text{Av}_\beta(\Delta U)} = k_B T.$$

- Av_β is the average with respect to the canonical density $\exp(-\beta H(q,p))$
- Special cases of general result:

Proposition 6.1 *Let H be a Hamiltonian defined on the phase space \mathbb{R}^{2N_c} and suppose $\mathbf{G} : \mathbb{R}^{2N_c} \rightarrow \mathbb{R}^{2N_c}$ is a smooth (C^1) vector field with the following properties:*

- $0 < |\text{Av}_\beta(\mathbf{G} \cdot \nabla H)| < \infty$,
- $0 < \text{Av}_\beta(\nabla \cdot \mathbf{G}) < \infty$,
- $|\mathbf{G}e^{-\beta H}| < \infty$

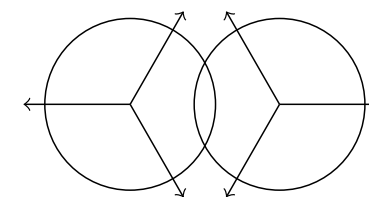
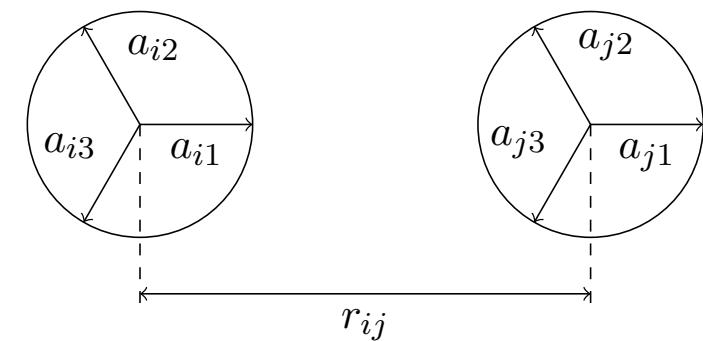
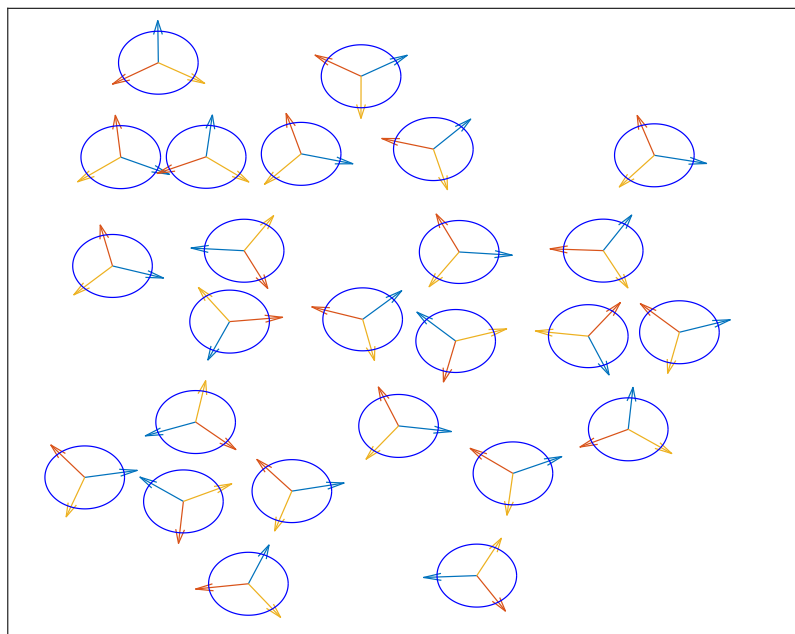
Then

$$k_B T = \frac{\text{Av}_\beta(\mathbf{G} \cdot \nabla H)}{\text{Av}_\beta(\nabla \cdot \mathbf{G})}.$$

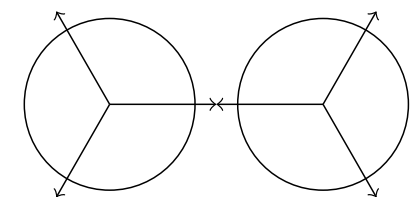
[Leimkuhler, Matthews '15]

The Mercedes Benz potential

- Model for water like particles
- Each particle has three arms originating from its center of mass. Thus particles are **anisotropic rigid bodies**.



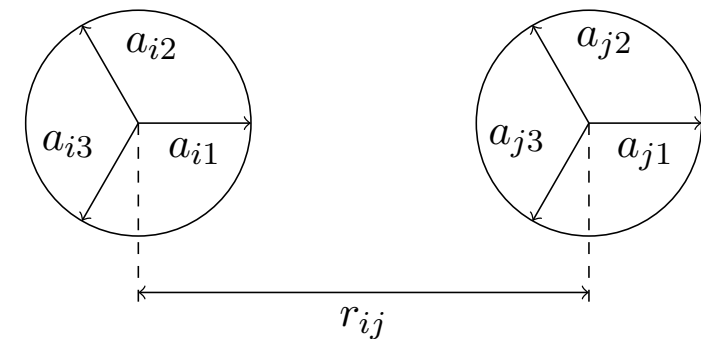
Double Bond



Single Bond

The Mercedes Benz potential

- Model for water like particles
- Each particle has three arms originating from its center of mass. Thus particles are **anisotropic rigid bodies**.



- MB pair potential:

$$\begin{aligned}
 V(x_i, x_j, Q_i, Q_j) &= 4\epsilon_{LJ} \left(\left(\frac{\sigma_{LJ}}{r_{ij}} \right)^{12} - \left(\frac{\sigma_{LJ}}{r_{ij}} \right)^6 \right) \\
 &\quad + \epsilon_{HB} G(r_{ij} - r_{HB}) \sum_{k,l=1}^3 G((Q_i^T a_{ik})^T u_{ij} - 1) G((Q_i^T a_{ik})^T u_{ij} + 1) \\
 &= V_{LJ}(x_i, x_j) + V_{HB}(x_i, x_j, Q_i, Q_j).
 \end{aligned}$$

$$G(x) = \exp \left(-\frac{x^2}{2\sigma_{HB}} \right)$$

center of mass

rotation matrix

Rigid bodies

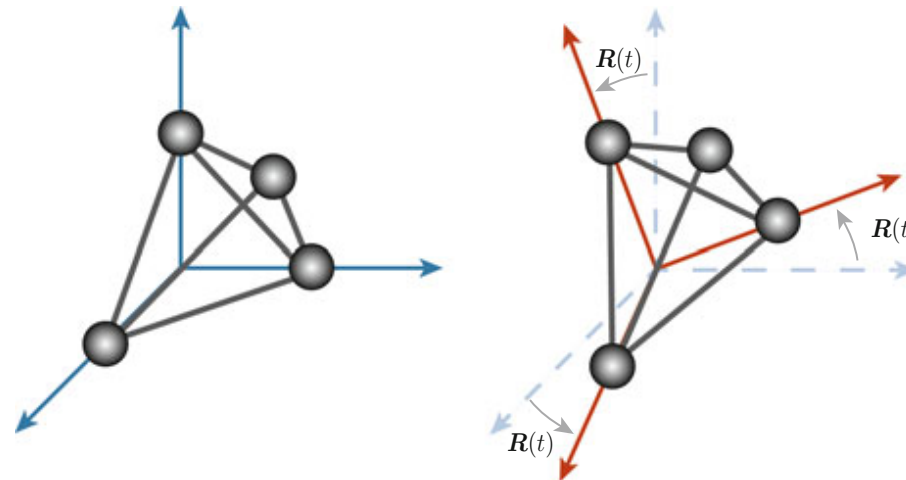


Fig. 4.7 A rigid rotation can be viewed as premultiplication by an orthogonal matrix Θ . In this way the orientation of the position vectors of points in the rigid body at any time t are described by a trajectory in the space of 3×3 orthogonal matrices, $\Theta(t)$, $t \in \mathbb{R}$

Two alternatives for representing rotational motion:

1. The rotation matrix $\Theta(t)$ describing the rotation of each particle in global reference frame is the dynamical quantity, and we solve for $\Theta(t)$. Need to enforce orthogonality constraints.

$$\begin{aligned} \frac{d}{dt} \mathbf{q}_{\text{cm}} &= \mathbf{p}_{\text{cm}}/M, \quad \frac{d}{dt} \Theta = \Pi R^{-1}, \\ \frac{d}{dt} \mathbf{p}_{\text{cm}} &= -\nabla_{\mathbf{q}_{\text{cm}}} U, \quad \frac{d}{dt} \Pi = -\nabla_{\Theta} U - \Theta \Lambda, \\ I &= \Theta^T \Theta. \end{aligned}$$

2. Use the orthogonality to perform reduction to angular momentum representation π with respect to body frame of reference.

The DLM rigid body integration algorithm

- Idea: Use splitting strategy
 - Solve for translational motion of center of masses,
 - Integrate torques due to rotational interactions
 - Solve Euler equations, recover rotation matrix

The DLM rigid body integration algorithm

1. “Kick”: update the momenta and angular momentum vectors with known forces and torques for a half step:

$$\mathbf{p}^k := \mathbf{p}^k + \frac{h}{2} \mathbf{F}^k, \quad \boldsymbol{\pi}^k := \boldsymbol{\pi}^k + \frac{h}{2} \boldsymbol{\tau}^k.$$

Forces

$$\mathbf{F}^k = -\frac{\partial U}{\partial \mathbf{q}^k},$$

2. “Drift”: step forward a whole step in positions

$$\mathbf{q}^k := \mathbf{q}^k + h\mathbf{p}^k.$$

Torques

$$\boldsymbol{\tau}^k = -\text{rot} \left([\boldsymbol{\Theta}^k]^T \frac{\partial U}{\partial \boldsymbol{\Theta}^k} \right).$$

3. “Spin”: update the rotational state of the k th rigid body by the sequence of steps

$$\begin{aligned} \boldsymbol{\Theta} &:= \boldsymbol{\Theta}^x \left(\frac{h}{2} \pi_1^k / I_1 \right); \boldsymbol{\pi}^k := \boldsymbol{\Theta} \boldsymbol{\pi}^k; \boldsymbol{\Theta}^k := \boldsymbol{\Theta}^k \boldsymbol{\Theta}^T \\ \boldsymbol{\Theta} &:= \boldsymbol{\Theta}^y \left(\frac{h}{2} \pi_2^k / I_2 \right); \boldsymbol{\pi}^k := \boldsymbol{\Theta} \boldsymbol{\pi}^k; \boldsymbol{\Theta}^k := \boldsymbol{\Theta}^k \boldsymbol{\Theta}^T \\ \boldsymbol{\Theta} &:= \boldsymbol{\Theta}^z \left(h \pi_3^k / I_3 \right); \boldsymbol{\pi}^k := \boldsymbol{\Theta} \boldsymbol{\pi}^k; \boldsymbol{\Theta}^k := \boldsymbol{\Theta}^k \boldsymbol{\Theta}^T \\ \boldsymbol{\Theta} &:= \boldsymbol{\Theta}^y \left(\frac{h}{2} \pi_2^k / I_2 \right); \boldsymbol{\pi}^k := \boldsymbol{\Theta} \boldsymbol{\pi}^k; \boldsymbol{\Theta}^k := \boldsymbol{\Theta}^k \boldsymbol{\Theta}^T \\ \boldsymbol{\Theta} &:= \boldsymbol{\Theta}^x \left(\frac{h}{2} \pi_1^k / I_1 \right); \boldsymbol{\pi}^k := \boldsymbol{\Theta} \boldsymbol{\pi}^k; \boldsymbol{\Theta}^k := \boldsymbol{\Theta}^k \boldsymbol{\Theta}^T \end{aligned}$$

$$\text{rot}(\mathbf{A}) = \text{skew}^{-1}(\mathbf{A} - \mathbf{A}^T).$$

The notation $\boldsymbol{\Theta}^x(\phi)$ refers to a rotation around the x -axis by an angle ϕ , that is

$$\boldsymbol{\Theta}^x(\phi) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos(\phi) & -\sin(\phi) \\ 0 & \sin(\phi) & \cos(\phi) \end{bmatrix},$$

$$\text{skew}(\boldsymbol{\omega}) \equiv \begin{bmatrix} 0 & -\omega_3 & \omega_2 \\ \omega_3 & 0 & -\omega_1 \\ -\omega_2 & \omega_1 & 0 \end{bmatrix}$$

with corresponding formulae for $\boldsymbol{\Theta}^y$ and $\boldsymbol{\Theta}^z$.

4. “Kick”: update the momenta and angular momentum vectors (as in Step 1) with a half step using new forces \mathbf{F}^k and torques $\boldsymbol{\tau}^k$ evaluated at the current \mathbf{q} , $\boldsymbol{\Theta}$.

Steps to perform the tutorial

- Read Mercedes Benz tutorial sheet
- Take the BAOAB integrator in MIST as a starting point
- Find a way to store the angular momentum of each particle in the system (hint: there is only one rotation in 2D)
- Modify the BAOAB integrator by incorporating the DLM algorithm
- Set up the force evaluations by calculating the derivatives of the MB potential

Thank you for your attention