

Molecular structures

Minimization, vibrational analysis, transition states

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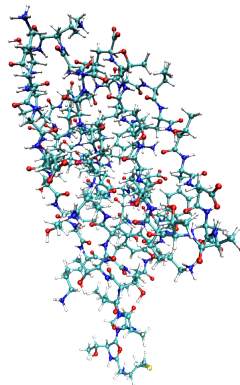
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Modeling of biomolecules

Potential energy surface: $E_{\text{el}} = E_{\text{el}}(\mathbf{R}_1, \dots, \mathbf{R}_N)$
(coordinates of atoms/nuclei $\mathbf{R}_1, \dots, \mathbf{R}_N$)

Approximations:

- Born–Oppenheimer approximation
 - separation of nuclei and electrons
 - E_{el} obtained for fixed positions of nuclei
- Classical description of nuclei
 - rather than quantum mechanics
for the motion of the nuclei
- Application of a force field
 - harmonic springs
 - point-charge electrostatics
 - ...



Energy with a force field

$$\begin{aligned} E_{\text{el}}(\mathbf{R}_1, \dots, \mathbf{R}_N) = & \frac{1}{2} \sum_i^{\text{bonds}} k_i (r_i - r_i^0)^2 + \frac{1}{2} \sum_i^{\text{angles}} k_i^\theta (\theta_i - \theta_i^0)^2 \\ & + \frac{1}{2} \sum_n^{\text{torsions}} \cos(n\omega_n - \gamma_n) \\ & + \sum_{I < J} \frac{1}{4\pi\epsilon_0} \frac{q_I q_J}{R_{IJ}} + \sum_{I < J} 4\epsilon_{IJ} \left(\left(\frac{\sigma_{IJ}}{R_{IJ}} \right)^{12} - \left(\frac{\sigma_{IJ}}{R_{IJ}} \right)^6 \right) \end{aligned}$$

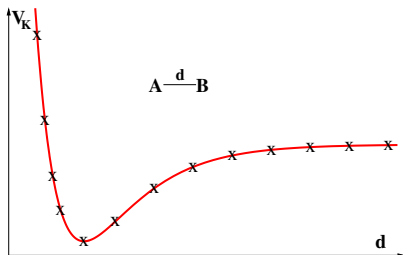
Potential energy surface

electronic energy is a function of coordinates of nuclei $\{\mathbf{R}_I\}$

$$E = E(\mathbf{R}_1, \dots, \mathbf{R}_N)$$

→ electronic energy defines the **potential energy surface** (PES)

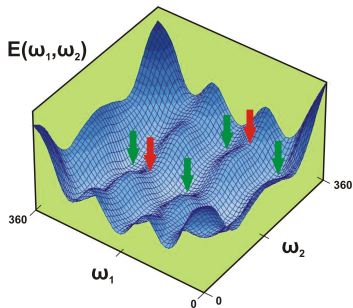
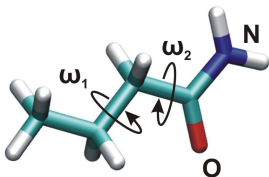
example: a diatomic molecule



→ all of the calculations only provide the PES point-wise

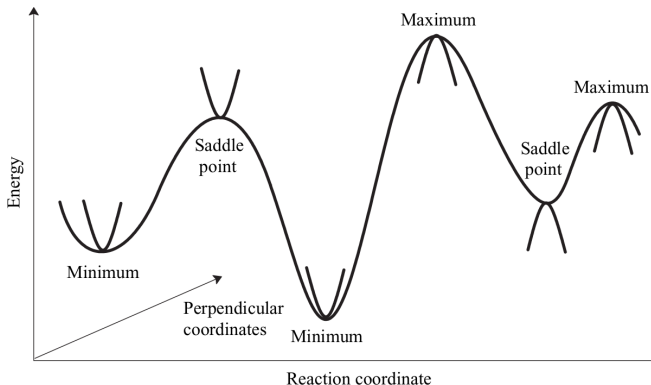
Sampling of the potential energy surface

chemically interesting:
stationary points
on the PES



- **minimum:**
‘equilibrium’ structure,
stable conformation of a molecule
- **saddle point of 1st order:**
transition state (TS),
point of maximum energy along the direction of a ‘reaction’

Potential energy surface

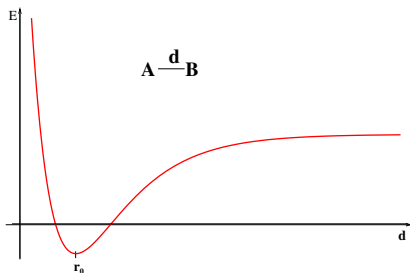


- **minimum:** any change of structure $\rightarrow$ increase of energy
- **saddle point of 1st order:** maximum along one coordinate, (reaction coordinate) $\times$ minimum along any other coordinate

Characterization of stationary points

How shall we find the interesting stationary points?

Example: Minimum of energy of a diatomic molecules



- condition for a stationary points: $\frac{dE}{dd} = 0$
- additional condition for a minimum: $\frac{d^2E}{dd^2} \geq 0$

Characterization of stationary points

How shall we find the interesting stationary points?

generally: $3N$ atom coordinates; $E = E(\mathbf{R}_1, \dots, \mathbf{R}_N)$

Condition for stationary points:

$$\text{Gradient: } \mathbf{g} = \nabla E = \left(\frac{dE}{dx_1}, \frac{dE}{dy_1}, \frac{dE}{dz_1}, \frac{dE}{dx_2}, \dots, \frac{dE}{dz_N} \right)^T = 0$$

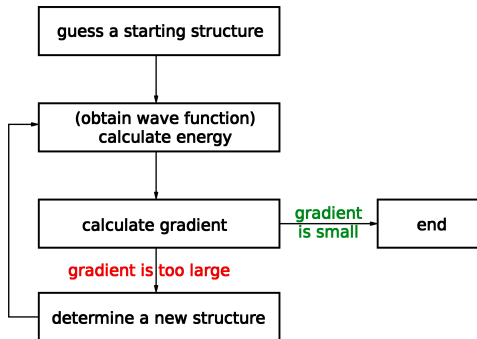
Second derivatives:

$$\text{Hessian: } \mathbf{H} = \begin{pmatrix} \frac{d^2E}{dx_1^2} & \frac{d^2E}{dx_1dy_1} & \frac{d^2E}{dx_1dz_1} & \frac{d^2E}{dx_1dx_2} & \cdots \\ \frac{d^2E}{dy_1dx_1} & \frac{d^2E}{dy_1^2} & \frac{d^2E}{dy_1dz_1} & \frac{d^2E}{dy_1dx_2} & \cdots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

- **Condition for minimum:** **all** eigenvalues of $\mathbf{H}$ are positive
- **Saddle point of 1st order:** **one** eigenvalue of $\mathbf{H}$ is negative, **all of the others** are non-negative

Geometry optimization = Energy minimization

→ search for a local minimum, starting from a suitable structure



- starting structure?
 - depends on the chemical problem
- convergence criterion? e.g.

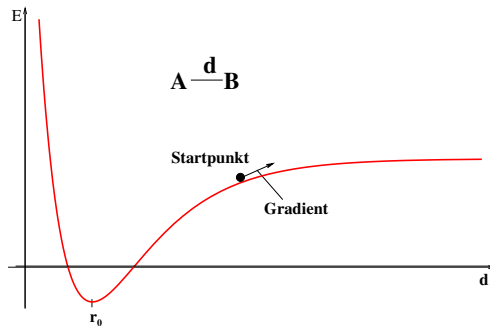
$$|\mathbf{g}| = \sqrt{\sum_{I,\alpha} g_{I,\alpha}^2} < t$$

- how shall we determine the new structure?
- try to make as few steps as possible
(calculation of energy/gradients is expensive)
- avoid any calculation of Hessian
(that is even more expensive)

Geometry optimization = Energy minimization

How to make a step towards a minimum?

Example: geometry optimization of a diatomic molecule



- follow the negative of gradient: $R_{i+1} = R_i + \Delta R_i = R_i - \alpha g_i$
- how shall the step length α be determined?

Steepest descents optimization

- Step along the negative of gradient

$$\Delta \mathbf{R}_i = \alpha \mathbf{d}_i \quad \mathbf{d}_i = -\mathbf{g}_i$$

- Choice of the step length α ?

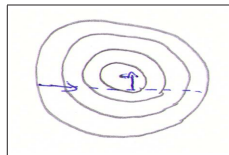
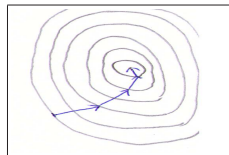
- too short $\rightarrow$ too many steps needed
- too long $\rightarrow$ overshoot the minimum

- 'line search':

choose α such that the energy
in direction of gradient $\mathbf{d}_i = -\mathbf{g}_i$
is minimized

– calculate the energy
at several points along a line

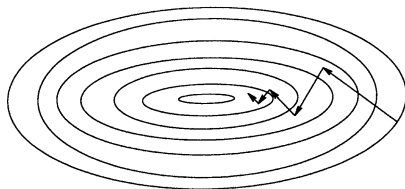
- Then, the convergence is guaranteed.



Conjugate gradient optimization

Problems of steepest descents

- too many steps in similar directions (in narrow valleys)
- convergence is getting slower when close to the minimum



Solution – conjugate gradient optimization

- make the step $\mathbf{d}_i$ orthogonal to *all* of the previous ones

$$\mathbf{d}_i = -\mathbf{g}_i + \beta_i \mathbf{d}_{i-1}$$

- various schemes: Fletcher–Reeves or Polak–Ribiere:

$$\beta_i^{\text{FR}} = \frac{\mathbf{g}_i^t \mathbf{g}_i}{\mathbf{g}_{i-1}^t \mathbf{g}_{i-1}} \quad \beta_i^{\text{PR}} = \frac{(\mathbf{g}_i - \mathbf{g}_{i-1})^t \mathbf{g}_i}{\mathbf{g}_{i-1}^t \mathbf{g}_{i-1}}$$

Newton–Raphson optimization

Taylor expansion of the PES around R_i :

$$E(\mathbf{R}) = E(\mathbf{R}_i) + \mathbf{g}_i \cdot (\mathbf{R} - \mathbf{R}_i) + \frac{1}{2}(\mathbf{R} - \mathbf{R}_i)^t \cdot \mathbf{H}_i \cdot (\mathbf{R} - \mathbf{R}_i) + \dots$$

minimum of this Taylor expansion up to the 2nd order:

$$\frac{dE}{d\mathbf{R}} = \mathbf{g}_i + \mathbf{H}_i \cdot (\mathbf{R} - \mathbf{R}_i) = 0 \rightarrow \mathbf{H}_i \cdot (\mathbf{R} - \mathbf{R}_i) = -\mathbf{g}_i$$

Newton–Raphson optimization:

calculate the step toward the minimum from that expression:

$$\Delta \mathbf{R}_i = -\mathbf{H}_i^{-1} \cdot \mathbf{g}_i$$

(on a harmonic PES, this would lead to the minimum directly)

Problem: the calculation of Hessian in every step is expensive

Quasi-Newton optimization

Apply an approximated Hessian rather than the exact Hessian

- start e.g. with a unit matrix ($\rightarrow$ 1 steepest-descents step)
- other starting Hessians may be better
 - full calculation (just once), or from certain simple rules
- in every step, use the gradients $\mathbf{g}_i$
to improve the approximated Hessian $\tilde{\mathbf{H}}_i$,
then invert it for $\Delta \mathbf{R}_i = -\mathbf{H}_i^{-1} \cdot \mathbf{g}_i$
- usually converges quickly and reliably, at the cost of
storage for Hessian of $\mathcal{O}(N^2)$ and its inversion of $\mathcal{O}(N^3)$
- compare: CG is $\mathcal{O}(N)$ but converges more slowly
- a standard method in most quantum chemical / MD packages

Quasi-Newton optimization

various update algorithms are available

- Broyden–Fletcher–Goldfarb–Shanno (**BFGS**):

$$\tilde{\mathbf{H}}_{i+1} = \tilde{\mathbf{H}}_i + \frac{\Delta \mathbf{g}_i \otimes \Delta \mathbf{g}_i^t}{\Delta \mathbf{g}_i^t \cdot \Delta \mathbf{R}_i} - \frac{\tilde{\mathbf{H}}_i \cdot \Delta \mathbf{R}_i \otimes \Delta \mathbf{R}_i^t \cdot \tilde{\mathbf{H}}_i}{\Delta \mathbf{g}_i^t \cdot \tilde{\mathbf{H}}_i \cdot \Delta \mathbf{R}_i}$$

- symmetric and positive-definite
- minimizes the change in Hessian

- limited-memory BFGS (**L-BFGS**)

- propagate the inverse Hessian (not the Hessian itself)
 - the $\mathcal{O}(N^3)$ matrix inversion is eliminated
- do not store the Hessian or inverse Hessian in memory
 - rather, reconstruct the matrix on the fly
 - from $\Delta \mathbf{g}_i$ & $\Delta \mathbf{R}_i$ over a few (< 10) last steps
 - the $\mathcal{O}(N^2)$ storage requirement is eliminated

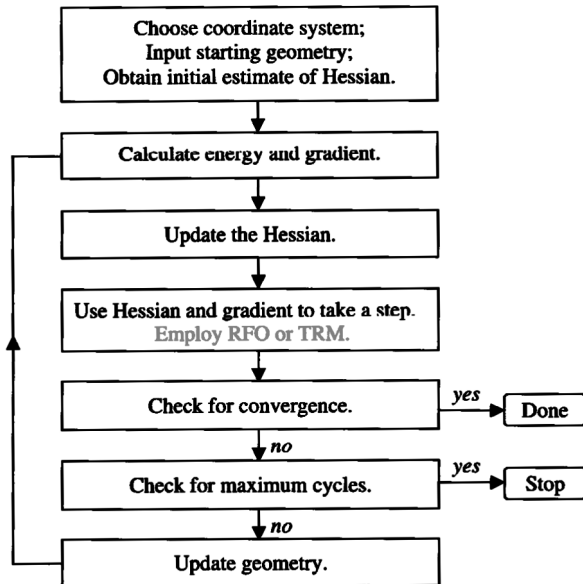
Quasi-Newton optimization

- q-N converges the better, the closer PES is to a quadratic form
- depends on **the choice of coordinate system** strongly

Possible choices of coordinate system

- cartesian coordinates
 - simple, but not adjusted to a ‘chemical’ problem
 - often slow convergence
- internal coordinates
 - use bond lengths, angles and torsional angles
 - often good convergence
 - but the definition of $3N - 6$ coordinates difficult
- redundant internal coordinates
 - use ‘too many’ internal coordinates
 - mostly good convergence
 - simple automatized definition is possible
- more complex coordinate systems possible

Quasi-Newton optimization



Summary: geometry optimization

- starting point: a chemically meaningful structure
- minimization procedures:
 - *steepest descents*: converges always, but slowly
 - better: *conjugate gradients*, *quasi-Newton* (e.g. *BFGS*)
 - all of these avoid the calculation of Hessian
- biomolecules – often very difficult to find true minima
- in the quantum chemistry
 - calculations mostly limited to a single minimum
 - starting point for the calculation of properties (spectra...)
- with the force field methods
 - starting point for MD – steepest descents is often good enough
 - pre-optimization for quantum chemical calculations

Limitations of the force field approximation

The parameters have to be determined / fitted

- difficult for certain / unusual elements (e.g., transition metals)

Conceptual limitations

- chemical bonds cannot be broken or created
- atom types are pre-determined
- atomic charges are pre-determined and constant
 - no change of electron density can be described

Electrons are not described explicitly, so no wave function

- only ground-state energy and forces are available
- no spectroscopic properties (interaction with light)
- no photochemistry (excited states)

Quantum chemistry vs. force fields

Quantum mechanics:

electronic wave functions: $\Psi_{\text{el}} = \Psi_{\text{el}}(\mathbf{r}_1, \dots, \mathbf{r}_N)$
(coordinates of electrons $\mathbf{r}_1, \dots, \mathbf{r}_N$)

solve the electronic Schrödinger equation

$$\hat{H}^{\{R\}} \Psi^{\{R\}} = E_{\text{el}}(\mathbf{R}_1, \dots, \mathbf{R}_N) \Psi^{\{R\}}$$

Quantum chemistry

- wave function theory (WFT)
- density functional theory (DFT)

→ calculation of energy is computationally intensive

Force field methods

- evaluate E_{el} 'directly', do not look for the electronic structure
- calculation of energy is very quick and efficient