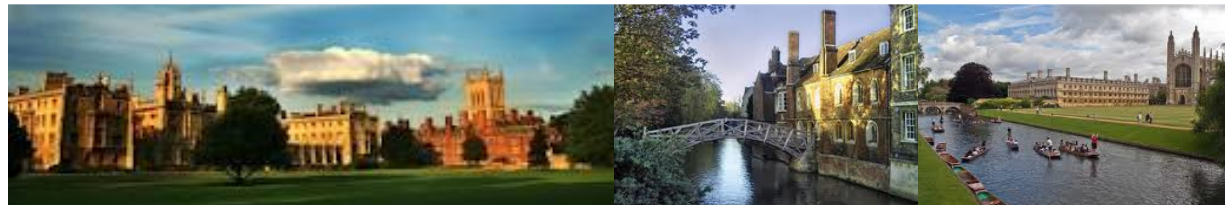




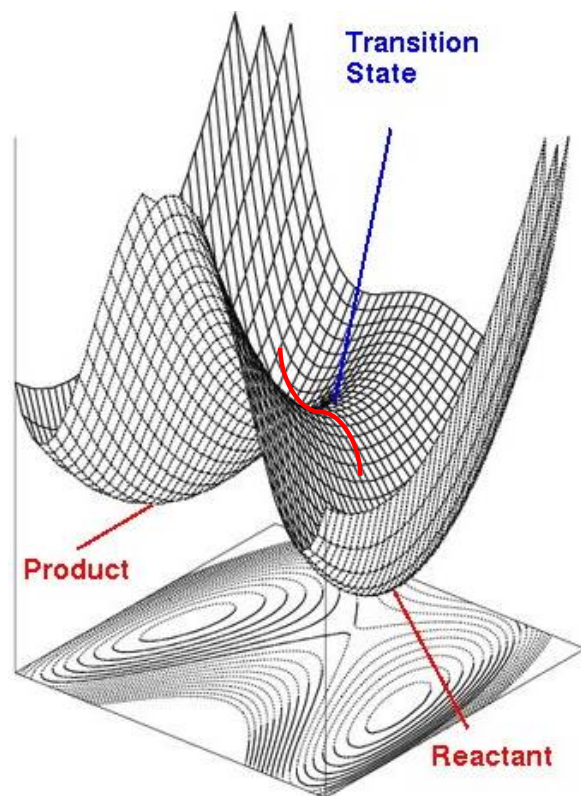
Forces and Structure Relaxation

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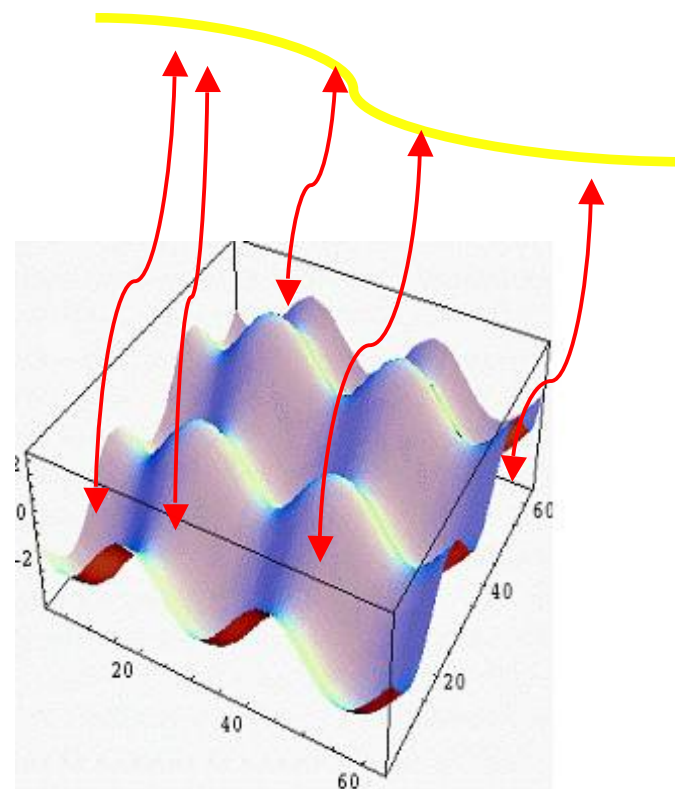


Potential Energy Surface (PES)



Optimizations and Phonons:

- We move on the PES
- Local vs global minima
- PES is harmonic close to minima



MD

- We move over the PES
- Good Sampling is required!!

DERIVATIVES OF TOTAL ENERGIES

- Many properties depend more on the derivatives of the energies, than the total energies themselves
- This can be computationally costly, and is susceptible to numerical noise
- We could get the derivatives by calculating the total energy at several points around each point, and do a numerical derivative
- Another approach is to use perturbation theory – in the form of the Hellman-Feynman theorem

The Hellman-Feynman theorem

Forces on atoms: $\mathbf{F} = -\nabla_{\mathbf{R}} \langle E \rangle$ where

$\langle E \rangle = \langle \Psi | H | \Psi \rangle$, if $|\Psi\rangle$ are normalised

$$\frac{\partial E}{\partial \lambda} = \frac{\langle \frac{\partial \Psi}{\partial \lambda} | H | \Psi \rangle}{\langle \Psi | H | \frac{\partial \Psi}{\partial \lambda} \rangle} + \langle \Psi | \frac{\partial H}{\partial \lambda} | \Psi \rangle +$$

$H|\Psi\rangle = E|\Psi\rangle$ and so

$$\frac{\partial E}{\partial \lambda} = \langle \Psi | \frac{\partial H}{\partial \lambda} | \Psi \rangle + E \frac{\partial}{\partial \lambda} \langle \Psi | \Psi \rangle$$

$$\frac{\partial E}{\partial \lambda} = \langle \Psi | \frac{\partial H}{\partial \lambda} | \Psi \rangle$$

Forces on the atoms

Hellman-Feynmann theorem:

$$F_I = -\frac{\partial \langle \psi | H | \psi \rangle}{\partial R_I} = -\cancel{\left\langle \frac{\partial \psi}{\partial R_I} | H | \psi \right\rangle} - \cancel{\left\langle \psi | H | \frac{\partial \psi}{\partial R_I} \right\rangle} - \left\langle \psi \left| \frac{\partial H}{\partial R_I} \right| \psi \right\rangle$$

Using an atomic-like basis:

$$\psi(\mathbf{r}) = \sum_{\mu} c_{\mu} \phi_{\mu}(\mathbf{r})$$

$$\begin{aligned} F_I &= -\frac{\partial \langle \psi | H | \psi \rangle}{\partial R_I} = -\frac{\partial}{\partial R_I} \sum_{\mu\nu} c_{\mu} c_{\nu} \langle \phi_{\mu} | H | \phi_{\nu} \rangle \\ &= -\sum_{\mu\nu} \left[\cancel{\frac{\partial c_{\mu}}{\partial R_I} c_{\nu}} + c_{\mu} \cancel{\frac{\partial c_{\nu}}{\partial R_I}} \right] \langle \phi_{\mu} | H | \phi_{\nu} \rangle - \sum_{\mu\nu} c_{\mu} c_{\nu} \left\langle \phi_{\mu} \left| \frac{\partial H}{\partial R_I} \right| \phi_{\nu} \right\rangle - \\ &\quad - \sum_{\mu\nu} c_{\mu} c_{\nu} \left[\left\langle \frac{\partial \phi_{\mu}}{\partial R_I} | H | \phi_{\nu} \right\rangle + \left\langle \phi_{\mu} | H | \frac{\partial \phi_{\nu}}{\partial R_I} \right\rangle \right] \end{aligned}$$

Pulay forces

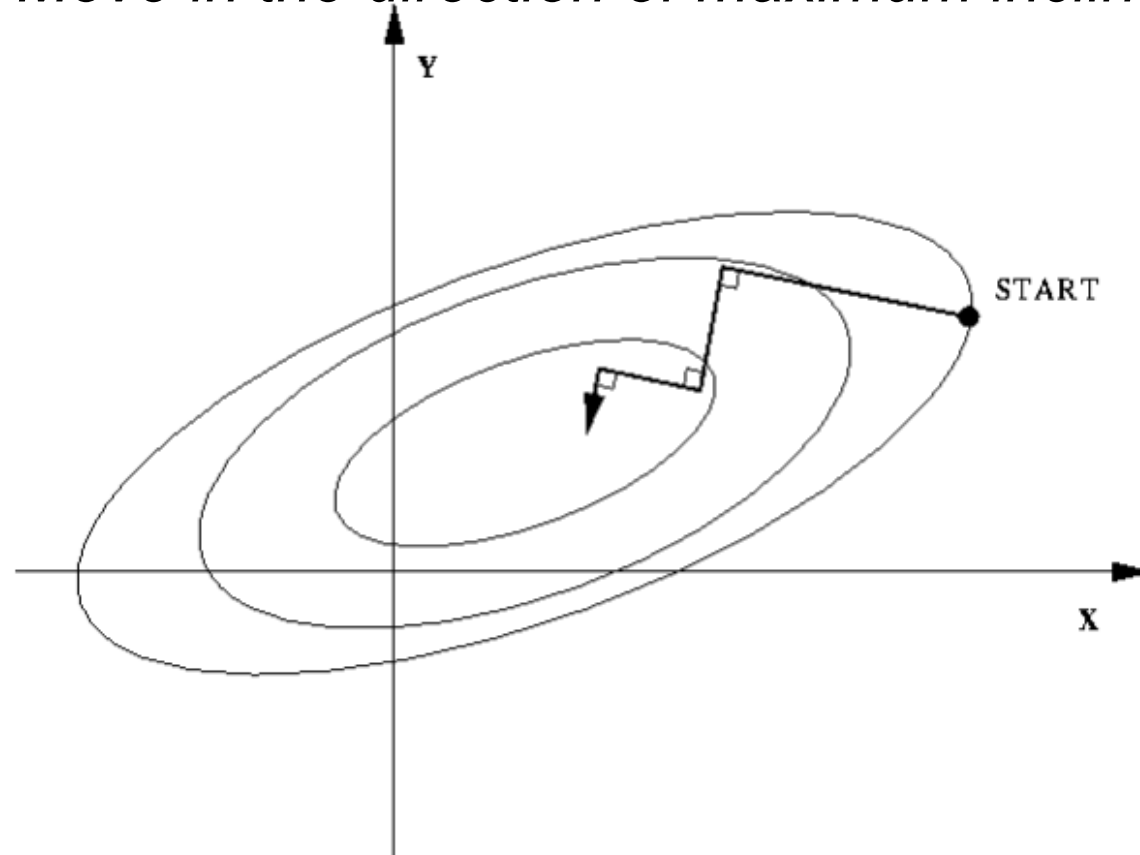
THE HELLMAN-FEYNMAN THEOREM AND DENSITY FUNCTIONAL THEORY

$$H_{\mathbf{R}} = -\frac{1}{2}\nabla_{\mathbf{r}}^2 + V_{\text{e-e}}(\mathbf{r}) + V_{\text{ion-e}}(\mathbf{r}, \mathbf{R}) + V_{\text{xc}}(\mathbf{r}) + V_{\text{ion-ion}}(\mathbf{R})$$

- The DFT Hamiltonian is parameterised in \mathbf{R}
- For the stresses, there are contributions from the kinetic energy and the Hartree terms
- We get a contribution from the ion-electron (pseudo)potential
- We get a contribution from the ion-ion Coulomb interaction (from the Ewald sum)
- While the total energy is correct to second order in the errors, the forces are only good to first order

Structural optimisation: Steepest descent

Move in the direction of maximum incline.

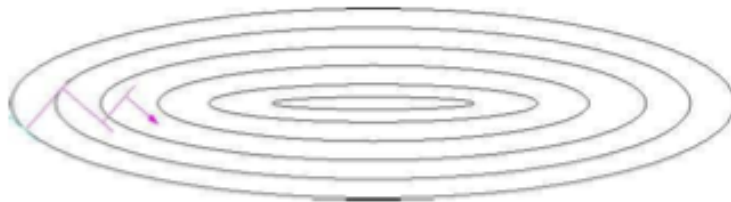


Repeats search directions

Structural optimisation: *minimum E vs atomic positions*

Follow forces on atoms (gradient of function)

STEEPEST DESCENTS



- This is the simplest approach:
 - take a downhill step along the local steepest gradient, and a trial step length
 - use line minimisation to find the optimal step length
- Advantages:
 - simple to implement, and robust
 - reliable – will find the minima eventually
- Disadvantages:
 - very slow to converge
 - can get stuck in a local minima

Theory for (local) geometry optimization

$$E(x + \delta x) = E(x) + G(x)\delta x + \frac{1}{2}H(x)\delta x^2$$



Gradients



Hessian

$$\delta x = -\alpha H(x)^{-1} G(x)$$



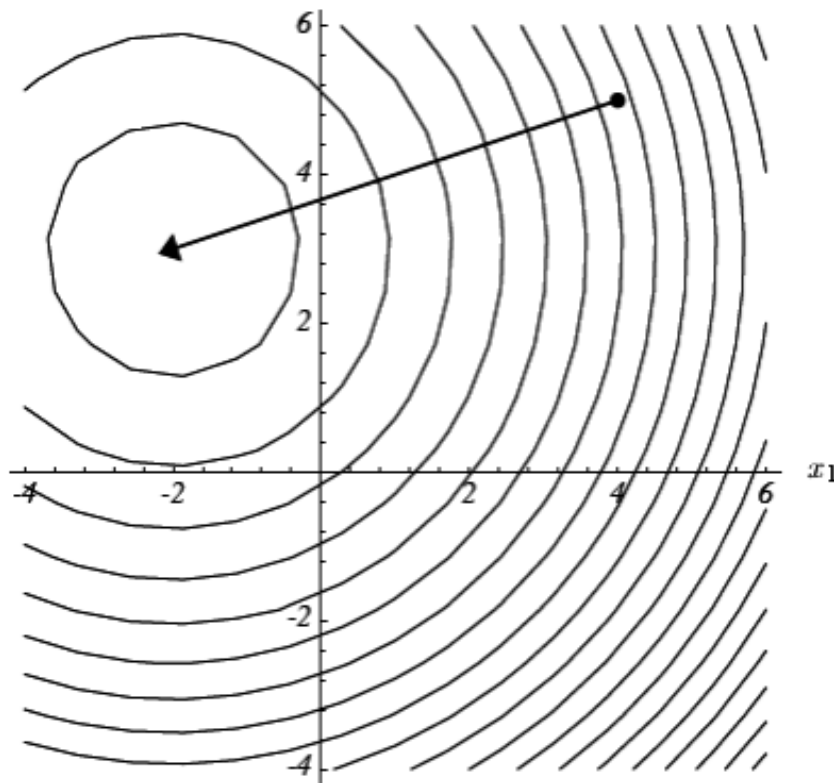
$\alpha = 1$ for quadratic region

Energy minimisation within a basin

Conditioning

Given eigenvalues of Hessian $H\vec{\chi}_i = \omega_i\vec{\chi}_i$

Condition number: $\kappa \propto \frac{\omega_i^{max}}{\omega_i^{min}}$ Determines convergence

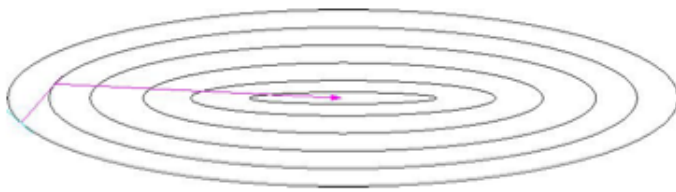


It converges on the first iteration If all ω are equal

Structural optimisation: *minimum E vs atomic positions*

Follow forces on atoms (gradient of function)

CONJUGATE GRADIENTS



- Advantages:
 - rapid convergence – in a quadratic energy landscape, one dof per iteration
 - low storage requirements
- This improves on steepest descents:
 - the gradient is constructed to be conjugate to all previous directions
 - does not undo previous minimisation
 - a line minimisation is performed
- Disadvantages:
 - more complex to implement than SD
 - Hessian not explicitly calculated
 - can get stuck in a local minima

Structural optimisation: *minimum E vs atomic positions*

Follow forces on atoms (gradient of function)

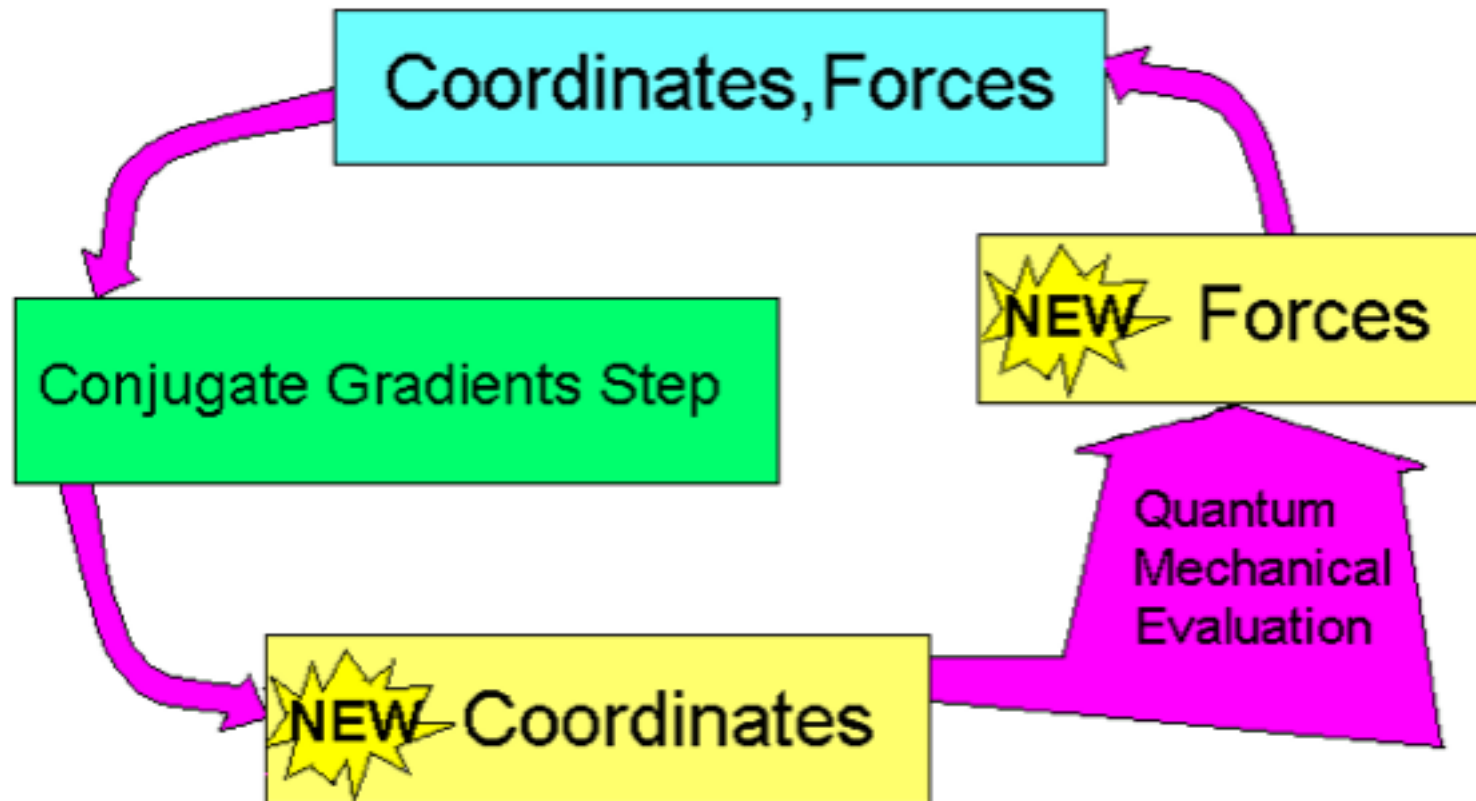
BROYDEN-FLETCHER-GOLDFARD-SHANNO – BFGS

$$\mathbf{A} = \begin{pmatrix} \frac{\partial^2 E}{\partial x_1 \partial x_1} & \cdots & \frac{\partial^2 E}{\partial x_1 \partial x_N} \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 E}{\partial x_N \partial x_1} & \cdots & \frac{\partial^2 E}{\partial x_N \partial x_N} \end{pmatrix}$$

$$\delta E = \frac{1}{2}(\mathbf{x} - \mathbf{x}_{\min})^T \cdot \mathbf{A} \cdot (\mathbf{x} - \mathbf{x}_{\min})$$

- Advantages:
 - convergence similar or better than CG
 - extra physical information is contained in the Hessian
- If we know the Hessian \mathbf{A} we can move from nearby the minimum straight to it
 - we don't know it, so we build up a guess using the BFGS algorithm
- Disadvantages:
 - complex to code
 - Hessian must be stored ($\# \text{ dof}^2$)
 - can get stuck in a local minima

Optimization (and MD) general basic Step



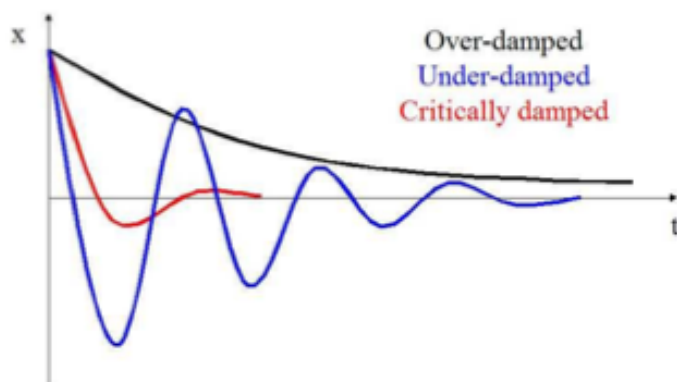
Optimization in SIESTA(1)

- *Set runtime to conjugate gradients:*
MD.TypeOfRun CG, Broyden
- *Set maximum number of iterative steps:*
MD.NumCGsteps 100
- *Optionally set force tolerance:*
MD.MaxForceTol 0.04 eV/Ang
- *Optionally set maximum displacement:*
MD.MaxCGDispl 0.2 Bohr

Structural optimisation: *minimum E vs atomic positions*

Follow forces on atoms (gradient of function)

DAMPED MOLECULAR DYNAMICS



- This improves on steepest descents:
 - use velocities as well as forces
 - start with $\mathbf{v} = \mathbf{0}$ and add damping term to forces $-\gamma\mathbf{v}$
 - adjust γ and time step to obtain optimal convergence
- Advantages:
 - simple to implement, robust and more efficient than SD
 - can use wavefunction extrapolation (or Car-Parinello)
- Disadvantages:
 - convergence rate depends on damping factor γ
 - can get stuck in a local minima

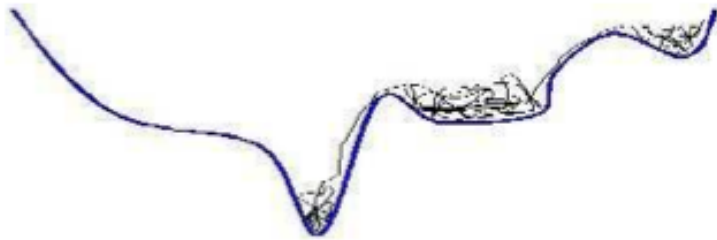
Structural optimisation:

global minimum E vs atomic positions

Follow forces on atoms (gradient of function)

SIMULATED ANNEALING

MD



- This is a stochastic method:
 - always accept steps that lower the energy, and sometimes accept upward steps, using a Boltzman distribution
 - slowly reduce temperature, and iterate to the groundstate
- Advantages:
 - very robust and reliable
 - reasonably immune to getting stuck in local minima
- Disadvantages:
 - incredibly slow convergence
 - the cooling rate must be carefully adjusted to avoid quenching into local minima
 - no guarantee that the true global minima will be found

Structural optimisation:

Variable cell

STRESS AND STRAIN

- The concept of forces is straightforward, but you can also take derivatives with respect to the crystal unit cell

$$\mathbf{h}' = (\mathbf{I} + \boldsymbol{\epsilon})\mathbf{h}$$

- The application of a *strain* changes the shape of the unit cell
- If we write the three unit cell vectors $\mathbf{a}, \mathbf{b}, \mathbf{c}$ as columns of a matrix \mathbf{h} the shape change is described by:

- The *stress* tensor $\boldsymbol{\sigma}$ is related to the strain tensor $\boldsymbol{\epsilon}$:

$$\sigma_{\alpha\beta} = \frac{1}{\Omega} \frac{\partial E}{\partial \epsilon_{\alpha\beta}}$$

where $\Omega = \mathbf{a} \cdot \mathbf{b} \times \mathbf{c}$ is the volume of the unit cell

Stress tensor and pressure

$$r_\alpha \rightarrow \sum_\beta \varepsilon_{\alpha\beta} r_\beta \quad \alpha, \beta = \{x, y, z\}$$

$$u_{i\alpha} \rightarrow \sum_\beta \varepsilon_{\alpha\beta} u_{i\beta} \quad \text{Strain tensor}$$

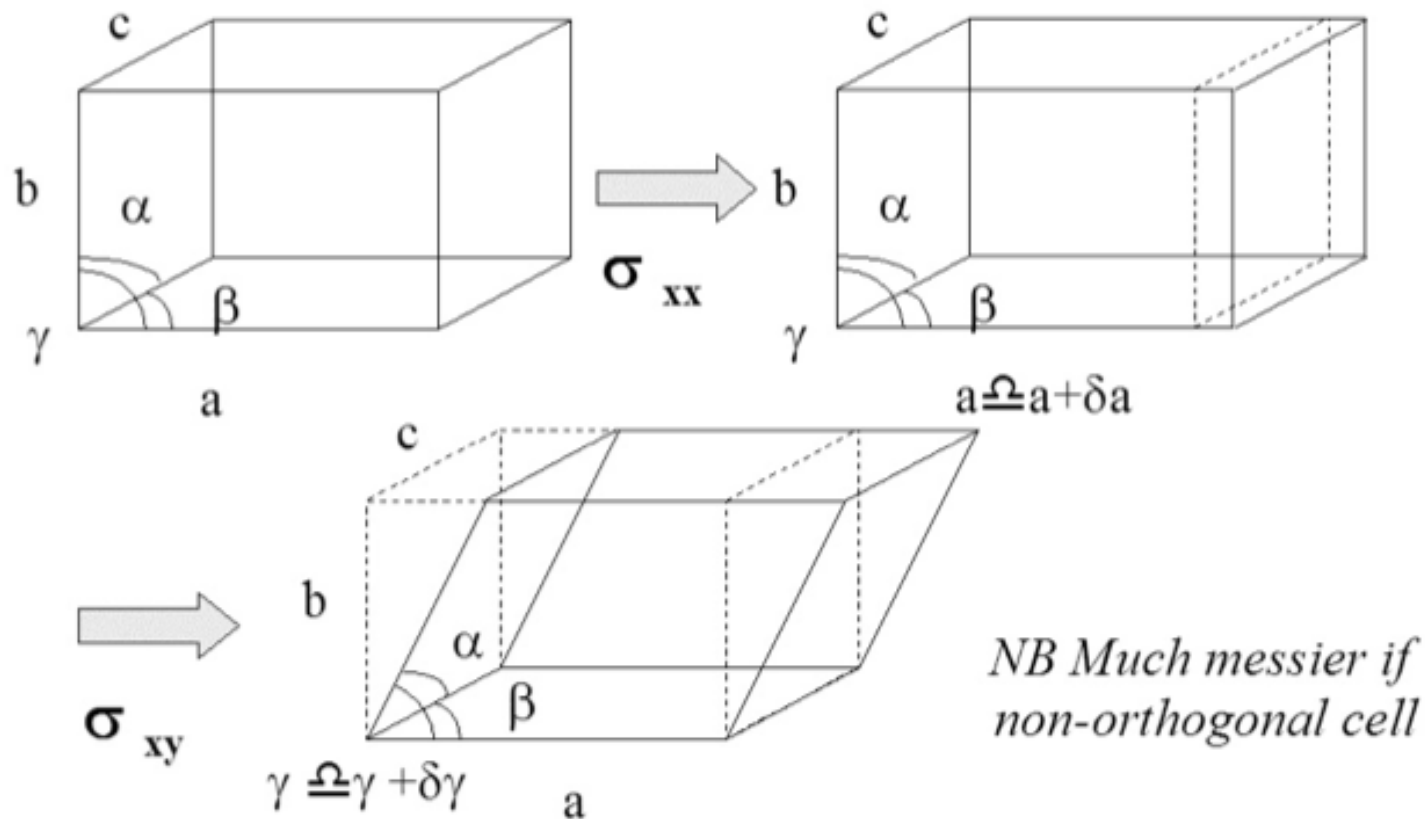
$$\sigma_{\alpha\beta} \equiv \frac{1}{\Omega} \frac{\partial E}{\partial \varepsilon_{\alpha\beta}} \quad \text{Stress tensor (Siesta)}$$

$$P = -\frac{1}{3} \sum_\alpha \sigma_{\alpha\alpha} \quad \text{Hydrostatic pressure}$$

$$P_{mol} = P - \frac{1}{\Omega} \sum_I \mathbf{R}_I \mathbf{F}_I \quad \text{Corrected pressure}$$

Structural optimisation: Variable cell

STRESS AND STRAIN



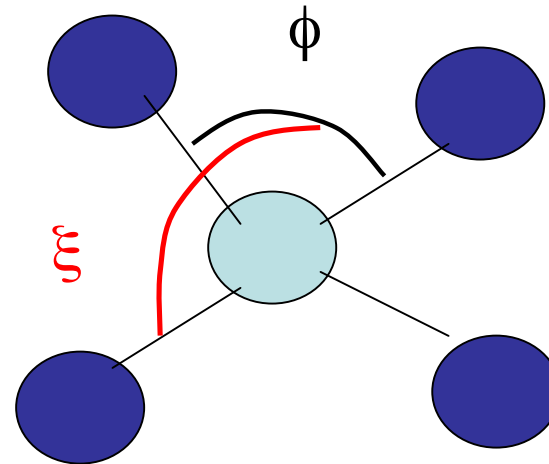
Optimizations in SIESTA(2)

By default optimisations are for a fixed cell

- *To allow unit cell to vary:*
MD.VariableCell true
- *Optionally set stress tolerance:*
MD.MaxStressTol 0.1 Gpa
- *Optionally set cell preconditioning:*
MD.PreconditionVariableCell 5.0 Ang
- *Set an applied pressure:*
MD.TargetPressure 5.0 GPa

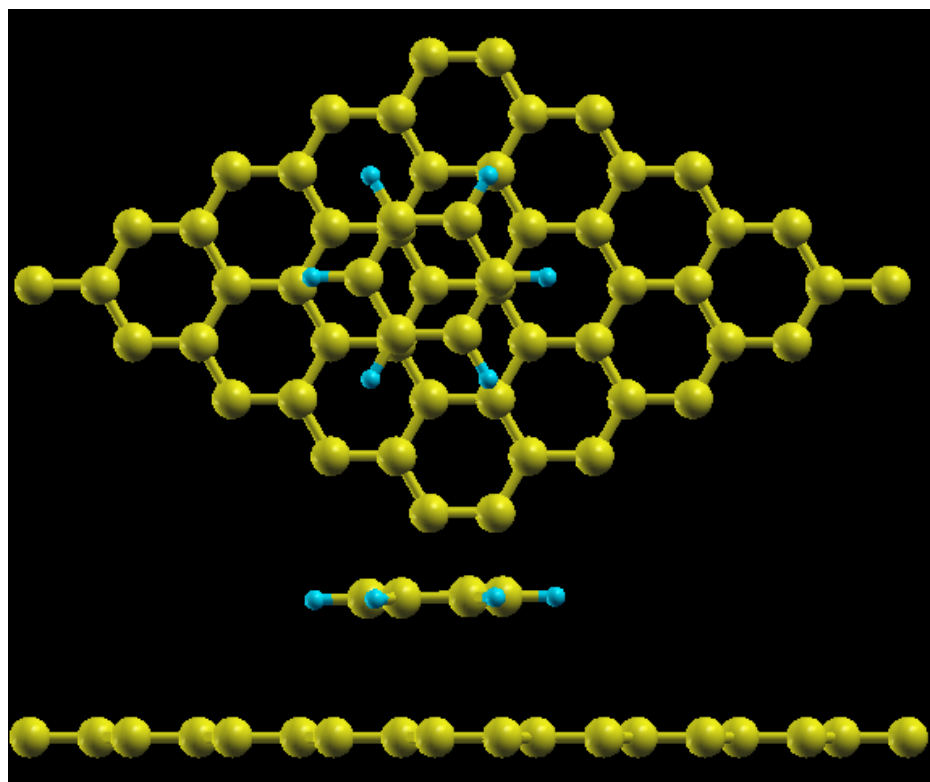
Z-Matrix coordinate format

- *Internal coordinates: Molecules represented by :*
 - Bond lengths χ_i
 - Bending angles ϕ_i
 - Dihedral angles ξ_i



Z-Matrix

- *Allows for mixing of generalised and Cartesian coordinates: Useful for constrained relaxations*



*Explore the PES by using
A relevant coordinate:
Useful for estimating
barriers*

Advice on optimizations in SIESTA(I)

Make sure that your MeshCutoff is high enough:

- *Mesh leads to space rippling*
- *If oscillations are large convergence is slow*
- *May get trapped in wrong local minimum*

```
siesta: Atomic forces (eV/Ang):
  1  -0.303027  -1.280971  0.567721
  2   2.853904   0.005572  0.953702
  3  -1.431055   2.487200  0.957536
  4  -1.355350  -2.348017  0.895960
-----
Tot  -0.235529  -1.136217  3.374919
-----
Max   2.853904
Res   1.538755  sqrt( Sum f_i^2 / 3N )
-----
Max   2.853904  constrained
```

✗

```
siesta: Atomic forces (eV/Ang):
  1   0.006491  -0.000001  -0.695690
  2   0.609572   0.000000   0.253077
  3  -0.309017   0.538586   0.252556
  4  -0.309017  -0.538586   0.252556
-----
Tot  -0.001972  -0.000001   0.062500
-----
Max   0.695690
Res   0.389268  sqrt( Sum f_i^2 / 3N )
-----
Max   0.695690  constrained
```

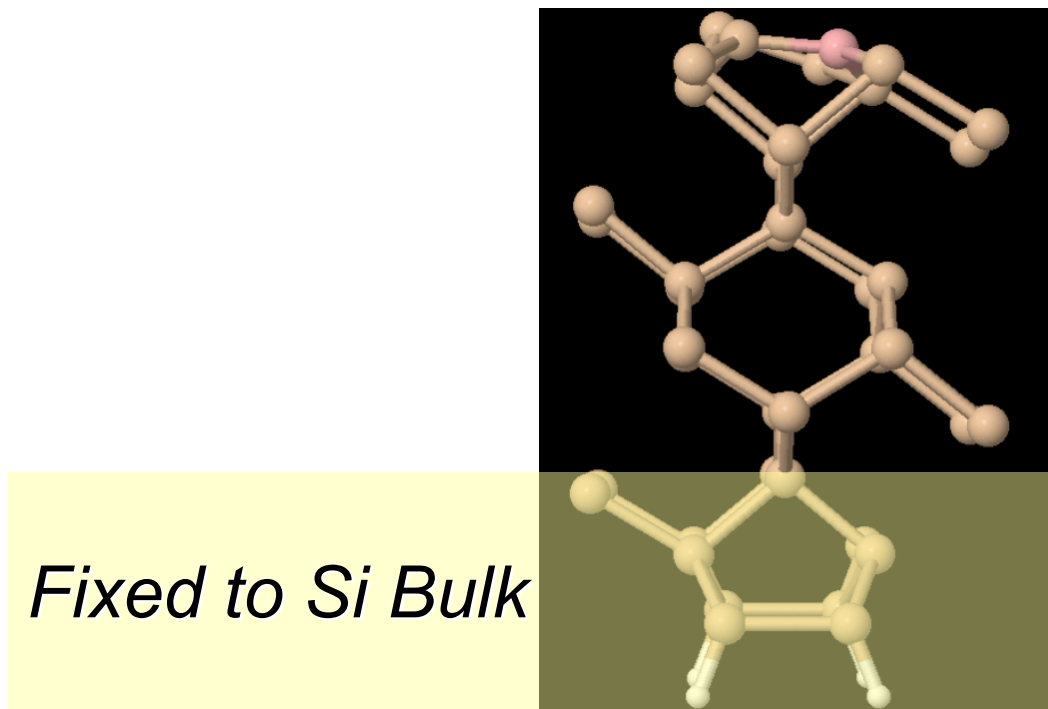
✓

Eggbox

Advice on Optimizations in SIESTA(II)

Ill-conditioned systems (soft modes) can slow down optimizations, very sensitive to mesh cutoff.

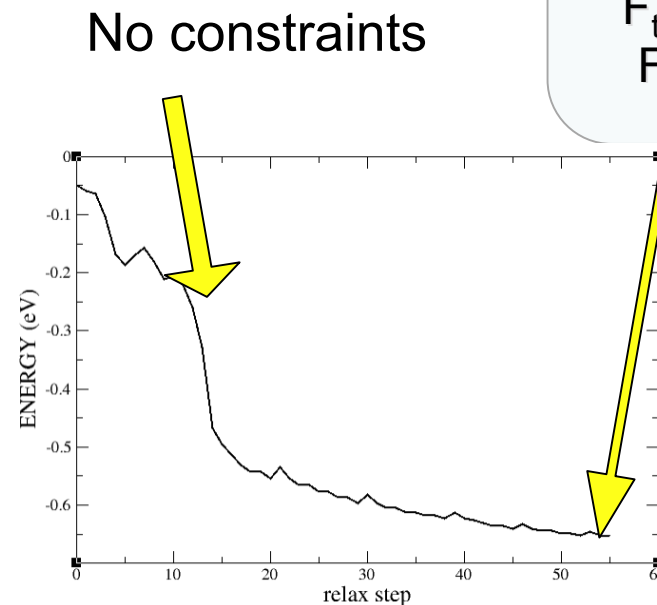
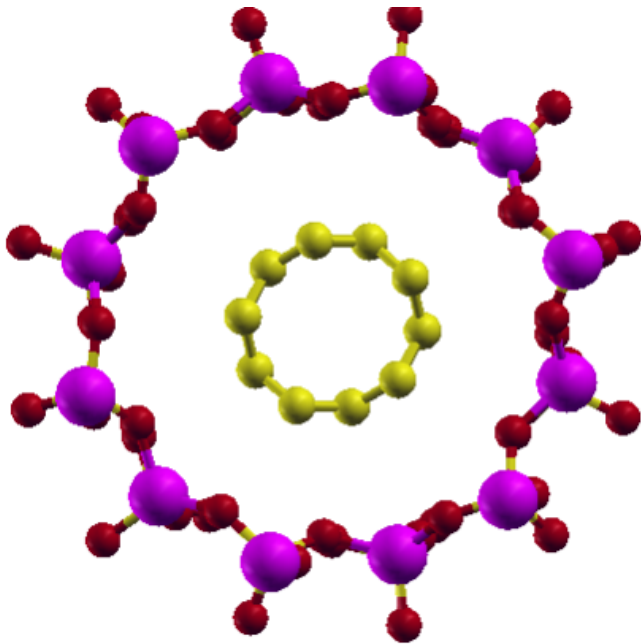
Use constraints when relevant.



Advice on Optimizations in SIESTA(III)

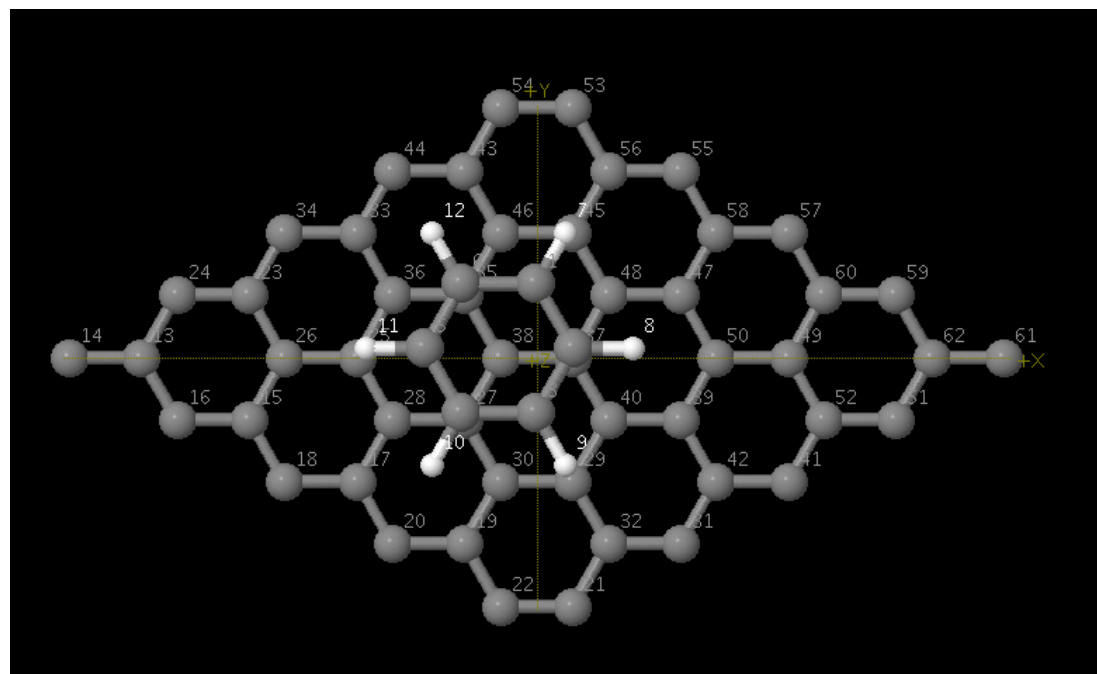
- *Decouple Degrees of freedom (relax separately different parts of the system).*

Look at the evolution of relevant physics quantities (band structure, E_f).



Fix the Zeolite,
Its relaxation is no
Longer relevant.

$$F_{\text{tube}} < 0.04 \text{ eV/\AA}$$
$$F_{\text{zeol}} > 0.1 \text{ eV/\AA}$$



%block Zmatrix

molecule

#N(l)	i	j	k	rlj	alji	tlkji	ifr	ifa	ift
1	0	0	0	0.00	1.396	zm1	0	0	1
1	1	0	0	CC	90.0	-60.0	0	0	0
1	2	1	0	CC	CCC	90.0	0	0	0
1	3	2	1	CC	CCC	0.0	0	0	0
1	4	3	2	CC	CCC	0.0	0	0	0
1	5	4	3	CC	CCC	0.0	0	0	0
2	1	2	3	CH	CCH	180.0	0	0	0
2	2	1	7	CH	CCH	0.0	0	0	0
2	3	2	8	CH	CCH	0.0	0	0	0
2	4	3	9	CH	CCH	0.0	0	0	0
2	5	4	10	CH	CCH	0.0	0	0	0
2	6	5	11	CH	CCH	0.0	0	0	0