



## Forces and Structure Relaxation

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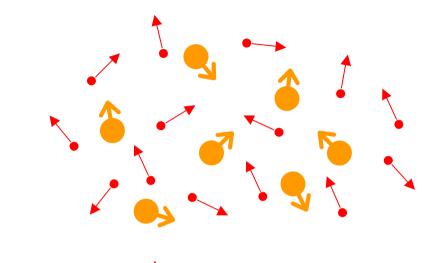
Nanogune, Ikerbasque & DIPC, San Sebastian, Spain Cavendish Laboratory, University of Cambridge





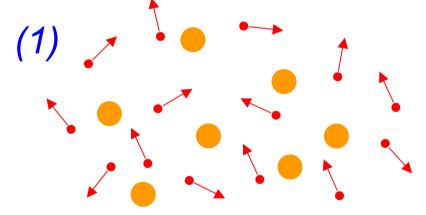


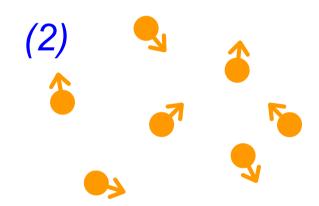
# Adiabatic decoupling



$$\frac{m_n}{m_e} >> 1$$

⇒Nuclei are much slower than electrons

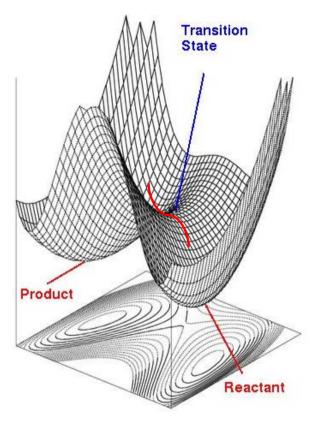




Quantum mechanics
Many electron problem:
Density Functional Theory

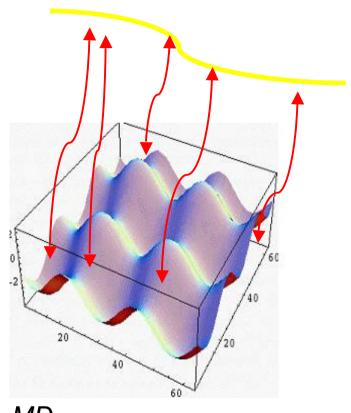
F = gradient of E: used to find minimum

## 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!!

# The Hellman-Feynman theorem

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

$$\mathbf{F} = - 
abla_{\mathbf{R}} \left\langle E 
ight
angle$$
 where

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

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

$$H|\Psi
angle=E|\Psi
angle$$
 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:

$$F_{I} = -\frac{\partial \langle \psi | H | \psi \rangle}{\partial R_{I}} = -\frac{\partial \psi}{\partial R_{I}} H | \psi \rangle - \frac{\partial \psi}{\partial R_{I}} - \frac{\partial \psi}{\partial R_{I}} - \frac{\partial \psi}{\partial R_{I}} | \psi \rangle$$

Using an atomic-like basis:

$$\begin{split} \psi(\mathbf{r}) &= \sum_{\mu} c_{\mu} \phi_{\mu}(\mathbf{r}) \\ F_{I} &= -\frac{\partial \left\langle \psi | H | \psi \right\rangle}{\partial R_{I}} = -\frac{\partial}{\partial R_{I}} \sum_{\mu\nu} c_{\mu} c_{\nu} \left\langle \phi_{\mu} | H | \phi_{\nu} \right\rangle \\ &= -\sum_{\mu\nu} \left[ \frac{\partial c_{\mu}}{\partial R_{I}} c_{\nu} + c_{\mu\nu} \frac{\partial c_{\nu}}{\partial R_{I}} \right] \left\langle \phi_{\mu} | H | \phi_{\nu} \right\rangle - \sum_{\mu\nu} c_{\mu} c_{\nu} \left\langle \phi_{\mu} \left| \frac{\partial H}{\partial R_{I}} \middle| \phi_{\nu} \right\rangle - \\ &- \sum_{\mu\nu} c_{\mu} c_{\nu} \left[ \left\langle \frac{\partial \phi_{\mu}}{\partial R_{I}} | H | \phi_{\nu} \right\rangle + \left\langle \phi_{\mu} | H | \frac{\partial \phi_{\nu}}{\partial R_{I}} \right\rangle \right] \end{split} \quad \text{Pulay forces}$$

# THE HELLMAN-FEYNMAN THEOREM AND DENSITY FUNCTIONAL THEORY

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

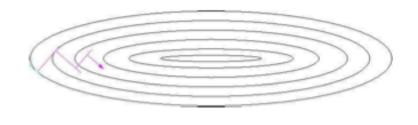
- The DFT Hamiltonian is parameterised
   in R
   contributions from the kinetic energy and the Hartree terms
- We get a contribution from the ionelectron (pseudo)potential
- We get a contribution from the ion- 

   While the total energy is correct to ion Couloumb interaction (from the second order in the errors, the forces are only good to first order

# Structural optimisation: minimum E vs atomic positions

Follow forces on atoms (gradient of function)

#### Steepest Descents



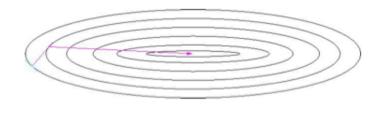
- Advantages:
  - simple to implement, and robust
  - reliable will find the minima eventually

- This is the simplest approach:
  - take a downhill step along the local steepest gradient, and a trial step • Disadvantages: length
  - use line minimisation to find the very slow to converge optimal step length
- - can get stuck in a local minima

# 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

## Theory for (local) geometry optimization

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

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

$$\alpha = 1 \text{ for quadratic region}$$

Energy minimisation within a basin

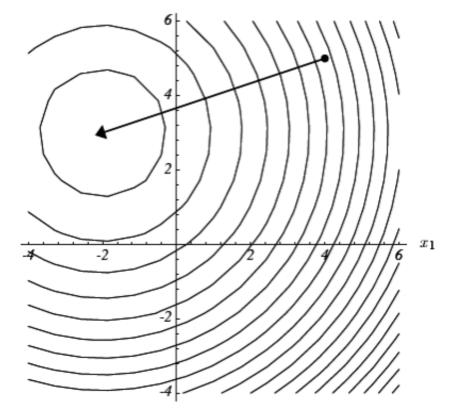
## Conditioning

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  $\omega$  are equal

## Structural optimisation: minimum E vs atomic positions

Follow forces on atoms (gradient of function)

Broyden-Fletcher-Goldfard-Shanno – BFGS

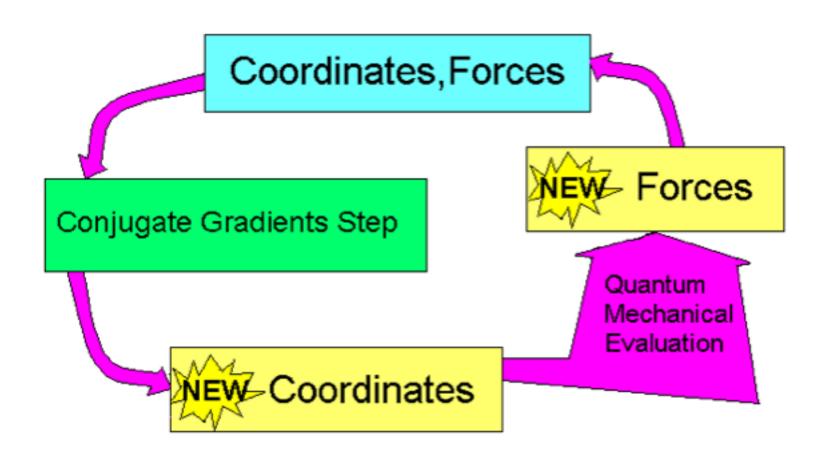
$$\mathbf{A} = \begin{pmatrix} \frac{\partial^2 E}{\partial x_1 \partial x_1} & \dots & \frac{\partial^2 E}{\partial x_1 \partial x_N} \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 E}{\partial x_N \partial x_1} & \dots & \frac{\partial^2 E}{\partial x_N \partial x_N} \end{pmatrix} \quad \begin{array}{l} \bullet \text{ Advantages:} \\ - \text{ convergence similar or better than} \\ \text{CG} \\ - \text{ extra_physical_information_is} \end{array}$$

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

- If we know the Hessian  ${\bf A}$  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 — can get stuck in a local minima

- extra physical information contained in the Hessian
- Disadvantages:
  - complex to code
- Hessian must be stored (# dof²)

## Optimization (and MD) general basic Step



## Optimization in SIESTA(1)

Set runtype 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

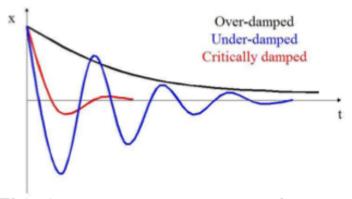
• Optionally set maximum displacement:

MD.MaxCGDispl 0.2 Bohr

# Structural optimisation: minimum E vs atomic positions

### Follow forces on atoms (gradient of function)

#### Damped Molecular Dynamics



#### Advantages:

- simple to implement, robust and more efficient than SD
- can use wavefunction extrapolation (or Car-Parinello)
- 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  $\gamma$  and time step to obtain optimal convergence

#### Disadvantages:

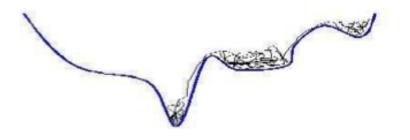
- convergence rate depends on damping factor  $\gamma$
- can get stuck in a local minima

# Structural optimisation: global minimum E vs atomic positions

Follow forces on atoms (gradient of function)

#### SIMULATED ANNEALING





- Advantages:
  - very robust and reliable
  - reasonably immune to getting stuck in local minima

- 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 goundstate

- 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
- The stress tensor σ is related to the strain tensor ε:

If we write the three unit cell vectors
 a, b, c as columns of a matrix h the shape change is described by:

$$\sigma_{lphaeta}=rac{1}{\Omega}rac{\partial E}{\partial\epsilon_{lphaeta}}$$

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

## Stress tensor and pressure

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

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

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

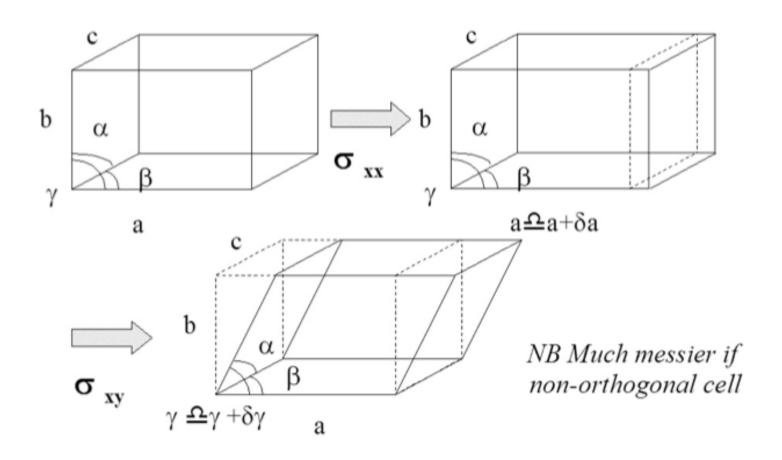
$$P = -\frac{1}{3} \sum_{\alpha} \sigma_{\alpha\alpha}$$
 Hydrostatic pressure

$$P_{mol} = P - \frac{1}{\Omega} \sum_{I} \mathbf{R}_{I} \mathbf{F}_{I}$$
 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:

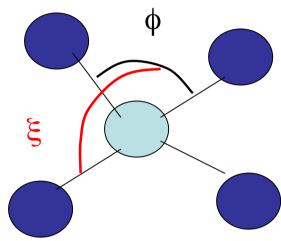
**MD.**TargetPressure

5.0 GPa

## Z-Matrix coordinate format

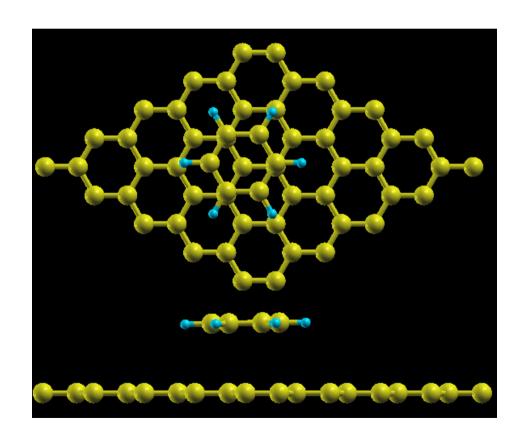
 Internal coordinates: Molecules represented by:

- Bond lengths  $\chi_i$
- Bending angles  $\phi_i$
- Dihedral angles 宾



## **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

siesto	a: Atomic fo	rces (eV/Ang)	:	siesta: Atomic forces (eV/Ang):				
1	-0.303027	-1.280971	0.567721	1	0.006491	-0.000001	-0.695690	
2	2.853904	0.005572	0.953702	2	0.609572	0.000000	0.253077	
3	-1.431055	2.487200	0.957536	3	-0.309017	0.538586	0.252556	
4	-1.355350	-2.348017	0.895960	4	-0.309017	-0.538586	0.252556	
Tot	-0.235529	-1.136217	3.374919	Tot	-0.001972	-0.000001	0.062500	
Max	2.853904	V201202505	-04904000 4 0940000	Max	0.695690			
Res	1.538755	sqrt( Sum f_i^2 / 3N )		Res	0.389268	sqrt( Sum f_i^2 / 3N )		
Max	2.853904	constraine	-d	Max	0.695690	constrained		



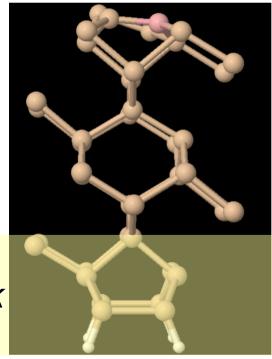


Eggbox ....

## Advice on Optimizations in SIESTA(II)

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

Use constraints when relevant.

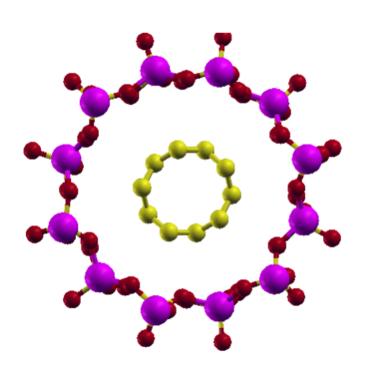


Fixed to Si Bulk

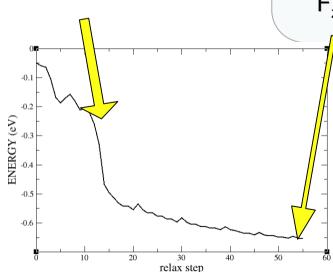
## Advice on Optimizations in SIESTA(III)

Decouple Degrees of freedom (relax separately different parts of the system). Look at the evolution of relevant physics quantities

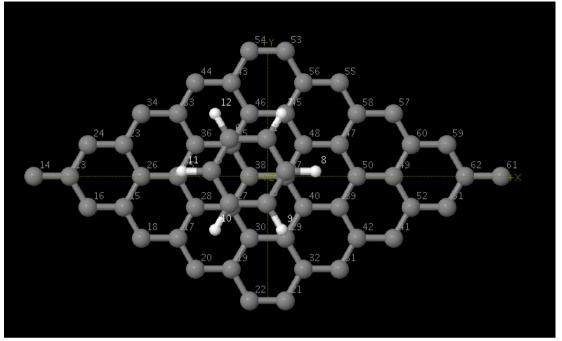
(band structure, Ef).



No constraints



Fix the Zeolite, Its relaxation is no Longer relevant.  $F_{tube}$ <0.04 eV/A  $F_{zeol} > 0.1 \text{ eV/A}$ 



9	6bloc	k Zn	natrix									
molecule												
7	#N(1)	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		