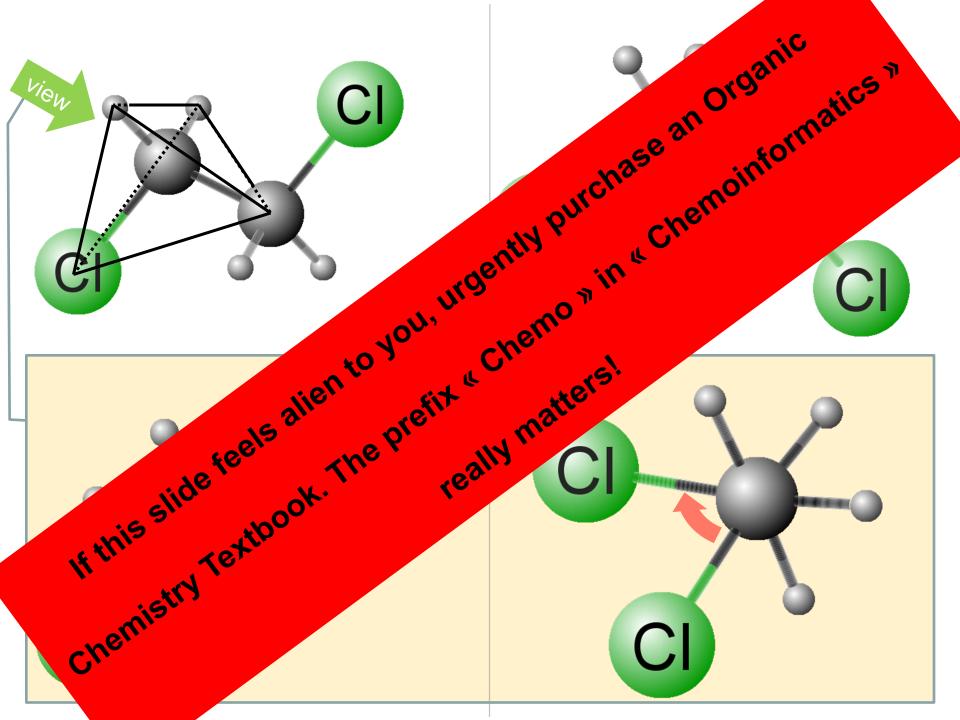
# **Conformational Sampling**

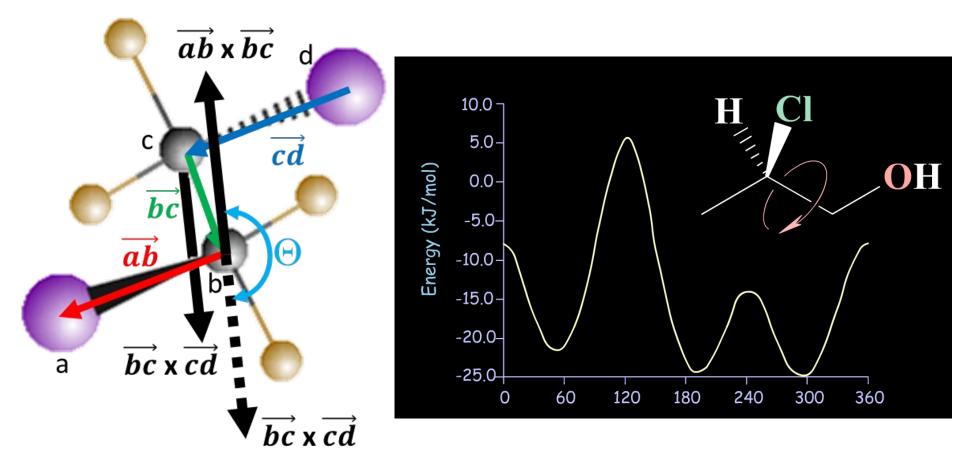
Dragos Horvath Laboratoire de Chemoinformatique – UMR 7140 dhorvath@unistra.fr

- Presentation Outline
  - The Basics: Molecules have Geometries!
    - Intramolecular energy: the Empirical Force Field
  - Sampling Methods: a brief overview
    - Molecular Dynamics: Walking like a molecule
    - Monte Carlo: Molecular Casino
    - Evolutionary Methods: in God/Darwin we trust!
  - Conclusions



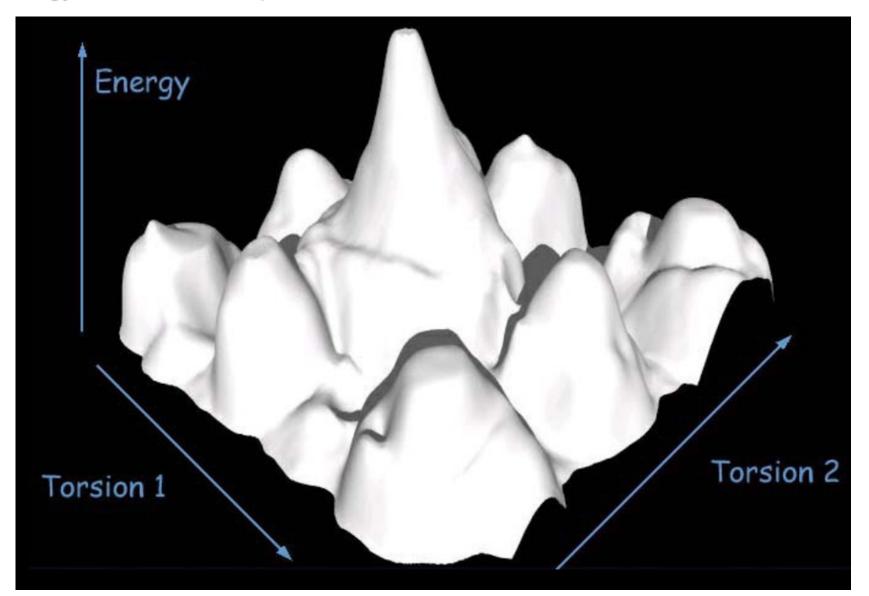
## Torsions : the gateway to conformational sampling

- Rotation around a bond impacts on interatomic distances, thus on energy!



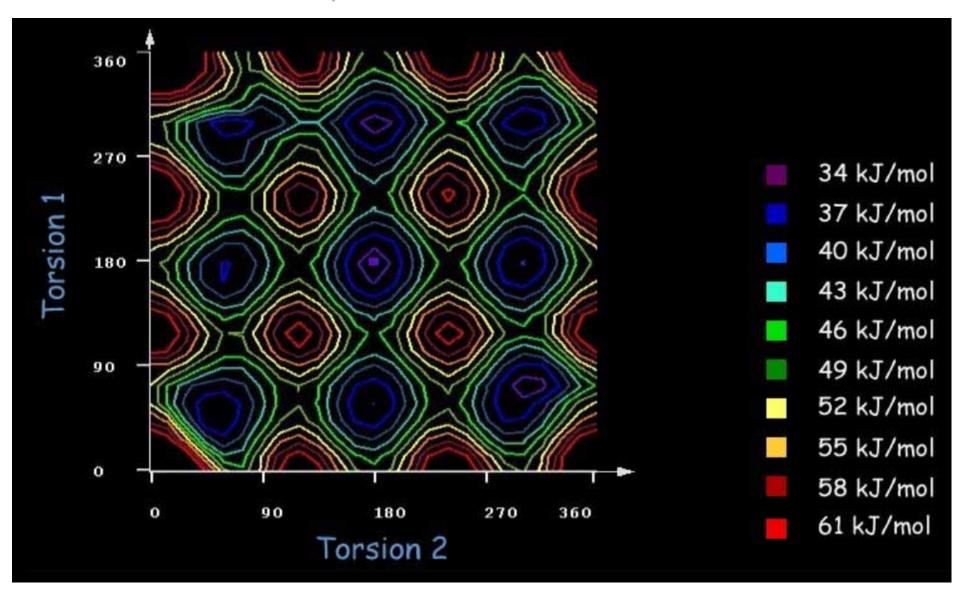
# Torsions : the gateway to conformational sampling

- Energy Surface with respect to two torsions....

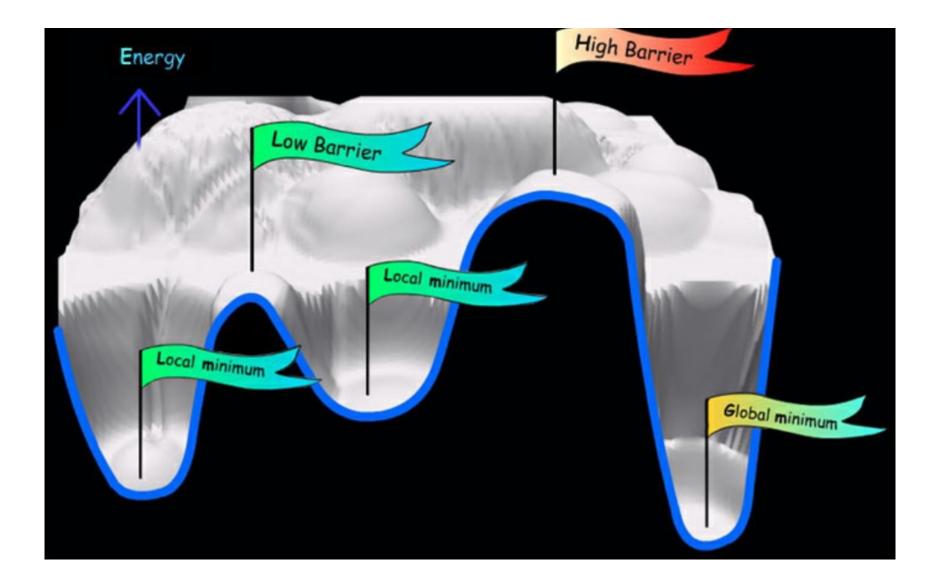


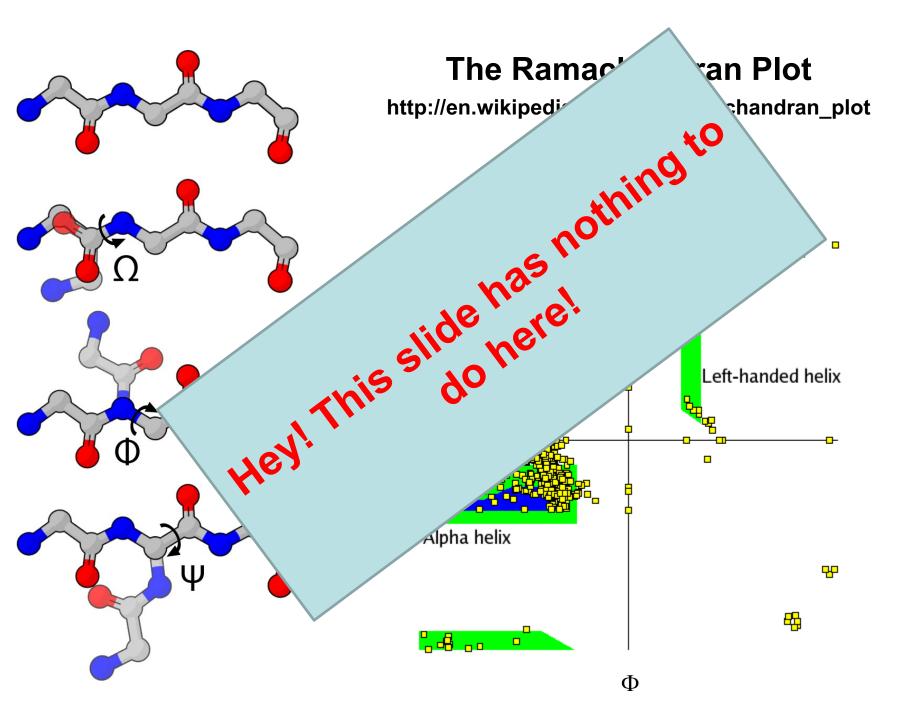
# Torsions : the gateway to conformational sampling

- Alternative Contour Plot representation



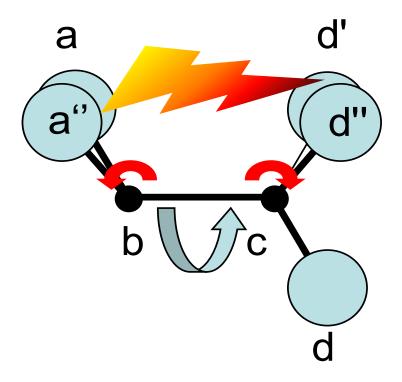
## Key points on the energy surface...





## Valence angle bending helps release some strain...

- Bond length oscillations mainly affect spectral properties



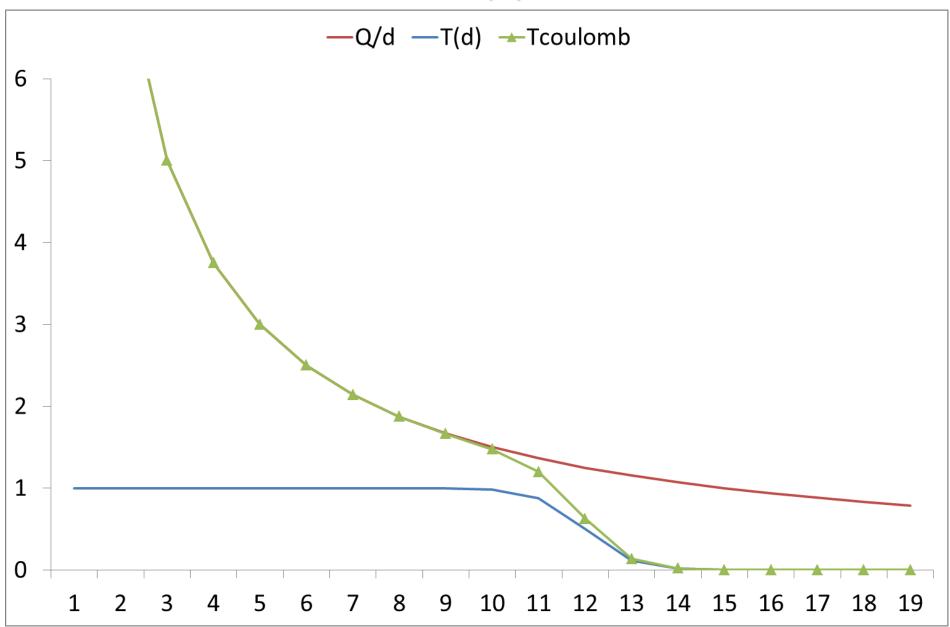
# Energy: function of internal coordinates, which depend on Cartesian x,y,z

Total Energy= f(bond1, bond2, bond3, bond4, bond5, bond6, bond7, bond8, bond8, bond9, bond10, bond11, bond12, bond13, bond14, bond15, bond24, bond25, bond26, bond27, bond28, bond29, bond30, bond31, bond32, bond33, bond34, bond35, bond36, bond37, bond38, bond39, bond40, bond41, bond42, valence2, valence3, valence4, valence ence9, valence10, valence11, valence ence16, valence17, HN valence18, valence znce23, valence24, valence25, valer ce29, valence30, valence31, valer ce35, valence36, valence37, vale ce41, valence42, =0 valence43, valen orsion3, torsion4, torsion5, torsion6 sion10, torsion11, torsion12, torsion12, in sions, in sions, in sions, in sion17, torsion18, torsion20, torsion21, torsion22, torsion23, torsion24, torsion25, torsion26, torsion27, torsion28, torsion29, torsion30, torsion31 torsion33, torsion34, torsion35, torsion36, torsion37, torsion38, torsion39, torsion40, torsion41, torsion42, torsion43, torsion44, torsion45

# Potential energy calculation is based on the Empirical Force Field (FF) approach

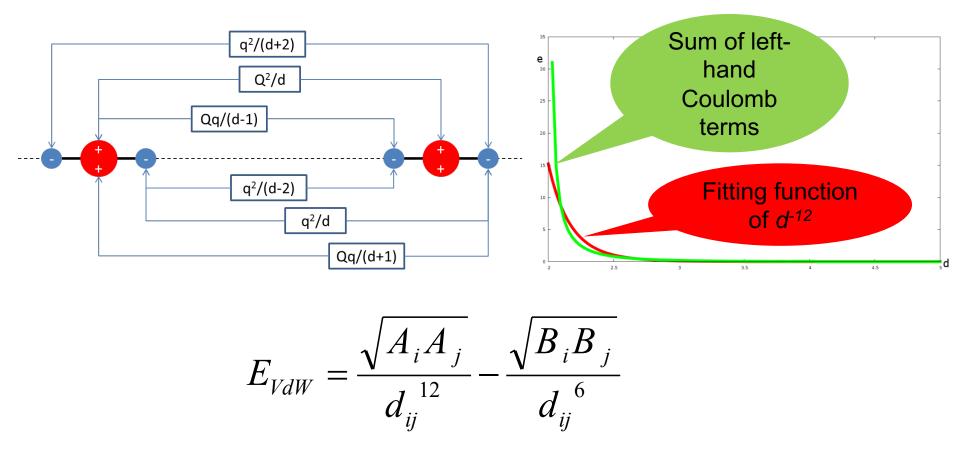
- Quantum chemical calculations are too time-consuming: atoms and their interactions are approximated as "classical" objects
- Atoms need to be "parameterized" in function of their chemical environment
- Covalent terms:
- bonds are modeled as harmonic springs. The energy required to stretch or compress a bond by  $\Delta b$  with respect to its natural length *b* is expressed as  $K_b \Delta b^2$ 
  - Valence angle bending modeled by harmonic potential  $K_{\phi} \Delta \phi^2$
- Atoms that are not directly bonded or do not form an angle interact "through space" by means of non-bonded interactions.
  - Electrostatics interactions based on partial charges
  - Continuum Solvent models
  - Van der Waals interactions
- Terms that should not be, but they're needed to make it work!
  - Torsional potentials, cross-terms, etc.

## Non-bonded interactions: (1) - Coulomb



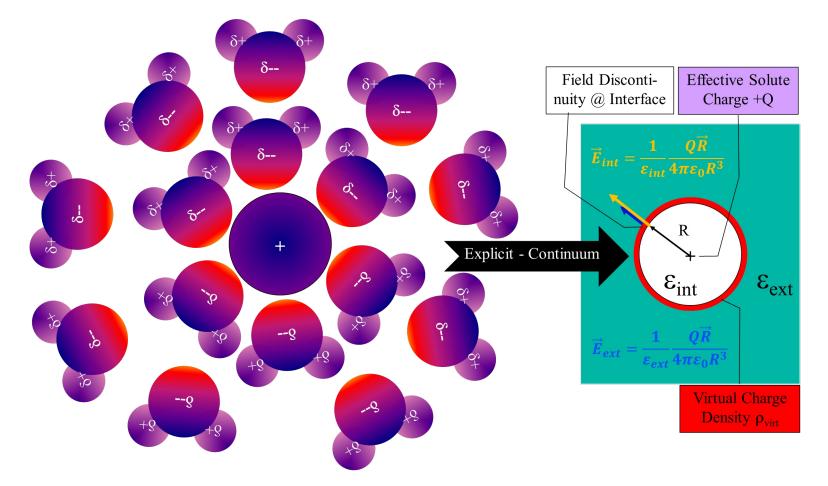
# Non-bonded interactions: (2) – van der Waals terms

- These are "cryptoelectrostatic" interactions:
  - a repulsive term acting at low distances, when electron spheres of nonbonded atoms overlap  $(d^{-12})$
  - An attractive term in *d*<sup>-6</sup> representing London dispersion terms (fluctuationinduced dipole-dipole interactions)

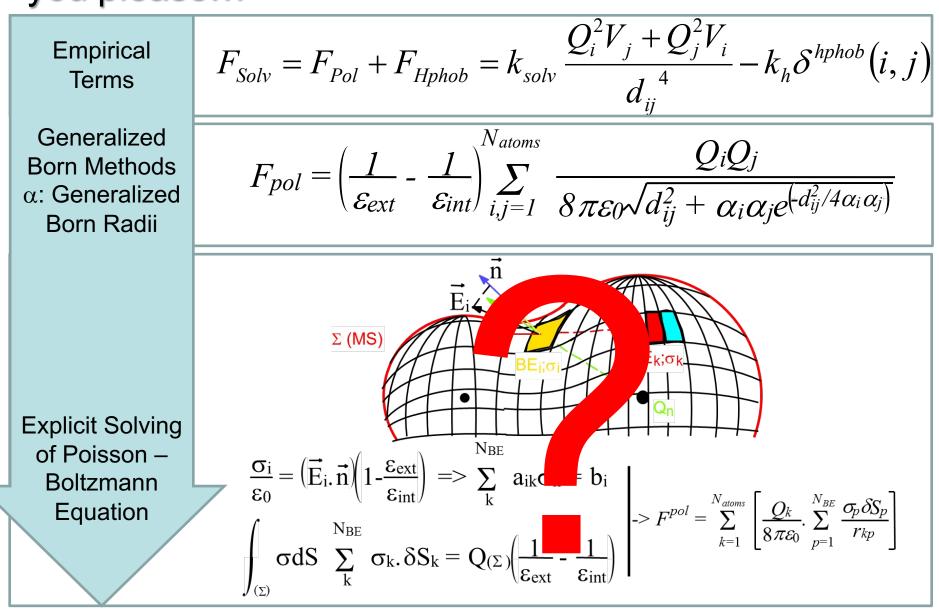


## Solvent effects...

- Explicit solvent molecules not only increase system size, but must be simulated in all their possible orientations with respect to the solute...
- Continuum solvent models assume dielectric effects to accurately capture this averaging over all possible solvent states...



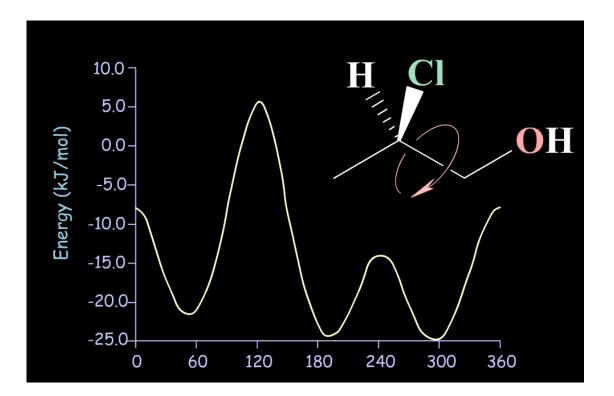
# Continuum solvent models may be as complex as you please...



Torsional terms – what for?

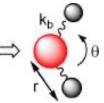
$$E_{tors} = k_t \left[ 1 + \cos(3\Theta - \Theta_0) \right]$$

- They must be introduced in order to correct for incompatibilities between long-range and short-range van der Waals contributions (N.Allinger, MM2)
- A set of FF parameters must be <u>consistent</u> do not expect individual terms to have any physical meaning (K. Rasmussen)

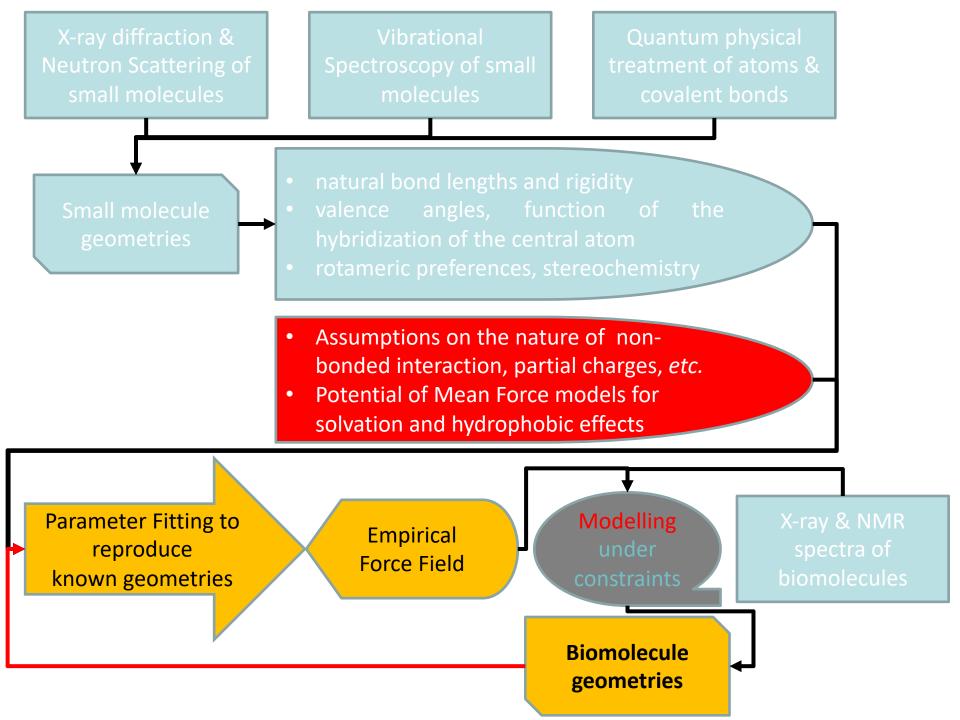


# Setting up a Force Field – where do all those parameters come from ?

- Few are directly issued from experimental observations:



- bond & angle deformation constants relate to IR vibration frequencies
- van der Waals parameters can be measured... for ideal gas atoms.
- Atomic partial charges from electronegativity equilibration, molecular orbital "collapsing".
- Most are <u>fitted</u>, making sure that force field simulations reproduce:
  - experimentally determined geometries & interconformational barriers
  - Quantum-chemically determined potential energy landscape



# Coarse-grain (CG) models for huge molecules....

- Replace standard fragments by some generic 'object' (bead):
  - What is the functional form of bead-bead interaction? Such "Potential of Mean Force" (PMF) should represent the average contact strength at inter-bead distance *d*.

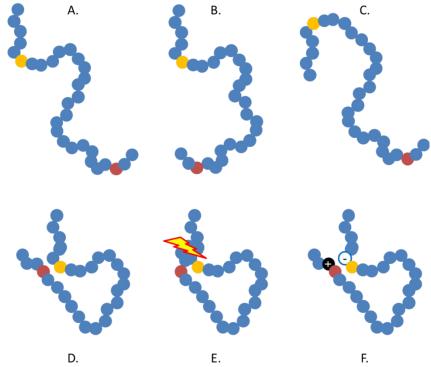
# $PMF_{CG}(\mathcal{A}) = -k_{B}T \ln Z_{d} = f\{E(\mathcal{A}), E(\mathcal{A}), E(\mathcal{A}), E(\mathcal{A}), E(\mathcal{A}), \dots\}$

# The back-engineering alternative: Knowledge-based potentials

If, in a structural data base, the red-yellow inter-particle distance is, on the average, *shorter* than expected (over the average of all possible chain arrangements – geometries like "D" seen more often than "normal"), then one may conclude that they **attract** each other

- O(d) = observed frequency @ d
- B(d) = baseline frequency @ d

$$\Rightarrow \exp\left[-\frac{E(d)}{k_bT}\right] = \frac{O(d)}{B(d)}$$



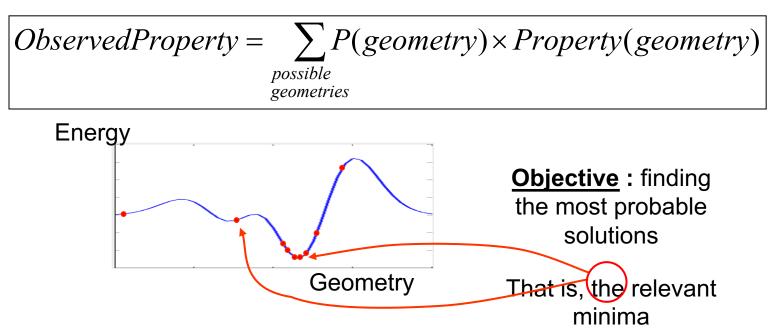
# The Question: Why care for conformational sampling?

 Because experimental properties of a molecule are given by the Boltzmann Average of properties of populated geometries

#### Boltzmann's probability distribution:

$$P(geometry \ of \ energy \ E) \sim \exp\left\{-\frac{E}{k_B T}\right\}$$

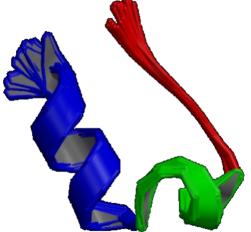
#### **Boltzmann Averaging:**



# By the way... what is a "Conformer"?

- A conformer is the set of geometries within the neighborhood of a local or global minimum energy geometry.
  - A geometry is a *point* in coordinate Phase Space,  $\vec{X} = (x_1, y_1, z_1, x_2, y_2, ..., x_N, y_N, z_N)$
  - A conformer is a Phase Space Domain C containing many similar geometries
- Therefore, a geometry is characterized by its energy, a conformer by its *partition function Z* and *free energy F*

$$Z(C) = \sum_{geometries \in C} \exp\left\{\frac{-E(geometry)}{k_B T}\right\}$$
$$F(C) = -k_B T \ln Z(C)$$



 At equal minimum depth *min E* over all geometries, the *broader energy well* (with more distinct low-energy geometries) will correspond to the experimentally more often seen conformer.

$$P(C) \sim \exp\left\{-\frac{F(C)}{k_B T}\right\}$$

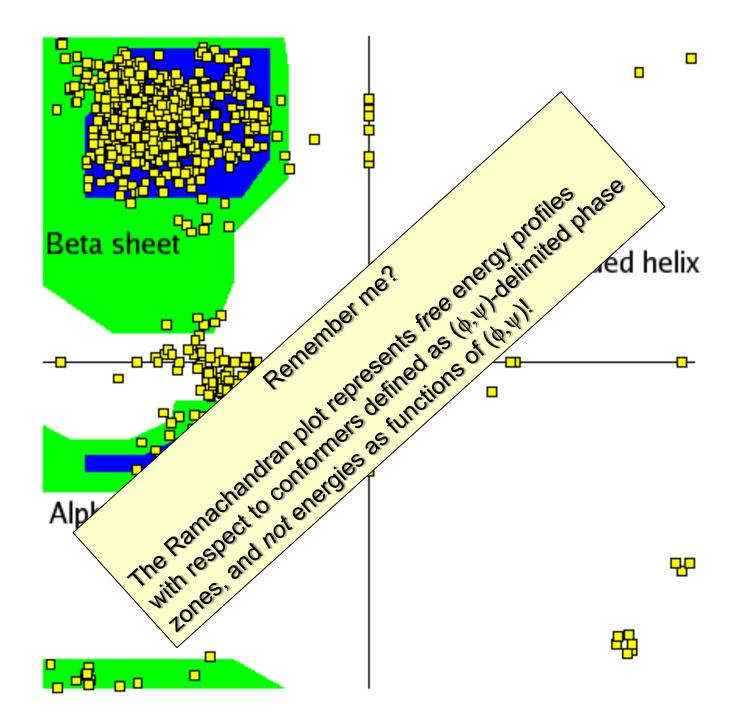
# ONE

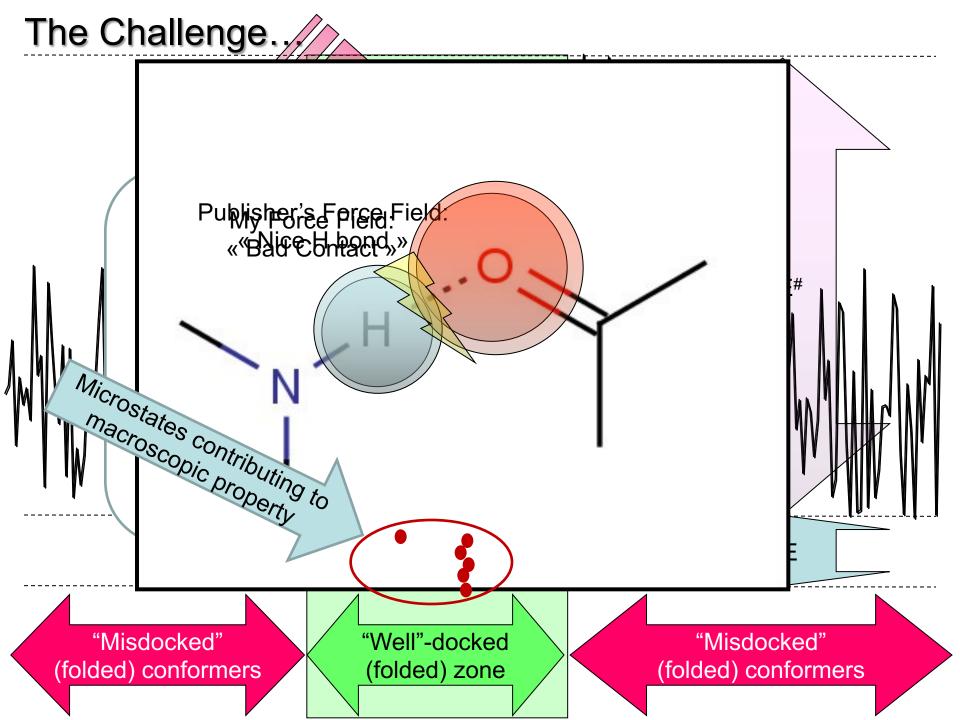
# THERMODYNAMICS AND STATISTICAL MECHANICS

#### 1.1 INTRODUCTION: THERMODYNAMICS AND STATISTICAL MECHANICS OF THE PERFECT GAS

Ludwig Boltzmann, who spent much of his life studying statistical mechanics, died in 1906, by his own hand. Paul Ehrenfest, carrying on the work, died similarly in 1933. Now it is our turn to study statistical mechanics. Perhaps it will be wise to approach the subject cautiously. We will begin by considering the simplest meaningful example, the perfect gas, in order to get the central concepts sorted out. In Chap. 2 we will return to complete the solution of that problem, and the results will provide the foundation of much of the rest of the book.

The quantum mechanical solution for the energy levels of a particle in a box (with periodic boundary conditions) is





- Presentation Outline
  - The Basics: Molecules have Geometries!
    - Intramolecular energy: the Empirical Force Field
  - Sampling Methods: a brief overview
    - Molecular Dynamics: Walking like a molecule
    - Monte Carlo: Molecular Casino
    - Evolutionary Methods: in God/Darwin we trust!

– Conclusions

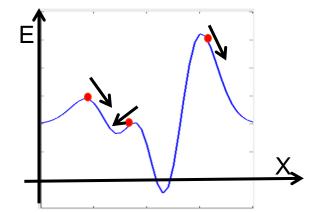
# Energy Minimization is only [the easy] part of the problem

- Given a starting geometry, deterministic algorithms allow the discovery of the adjacent local minimum
- Descent methods follow the local gradient

$$molecular \ geometry \ \vec{X} = (x_1, y_1, z_1, x_2, y_2, ..., x_N, y_N, z_N)$$

$$\nabla E(\vec{X}) = \left(\frac{\partial E}{\partial x_1}, \frac{\partial E}{\partial y_1}, \frac{\partial E}{\partial z_1}, \frac{\partial E}{\partial x_2}, \frac{\partial E}{\partial y_2}, ..., \frac{\partial E}{\partial x_N}, \frac{\partial E}{\partial y_N}, \frac{\partial E}{\partial z_N}\right)$$

$$iteratively: \ \vec{X}^{new} = \vec{X}^{curr} - s \nabla E(\vec{X}^{curr})$$



# Molecular Dynamics: walk the energy surface like the molecule does... following Newtonian dynamics

- Assume random atom velocities according to Maxwell's distribution at temperature T
- Calculate forces on atoms  $\vec{F} = -\nabla E(\vec{X}) = -\left(\frac{\partial E}{\partial x_1}, \frac{\partial E}{\partial y_1}, \frac{\partial E}{\partial z_2}, \frac{\partial E}{\partial x_2}, \frac{\partial E}{\partial y_2}, \dots, \frac{\partial E}{\partial x_N}, \frac{\partial E}{\partial z_N}, \frac{\partial E}{\partial z_N}\right)$
- Recalculate velocities at future time point:  $v(t+dt) = v(t) + F / m \times dt$
- Recalculate coordinates at future time point:

$$x(t+dt) = x(t) + \frac{v(t) + v(t+dt)}{2} \times dt$$

- Continue for the relevant time scale of the phenomenon to simulate... all while recalling that *dt* must be small enough to ensure that F≈constant during *dt*.
  - Bond stretching forces are the most rapidly varying they impose *dt*=10<sup>-15</sup> s

# But... a molecule is never alone: need to simulate the its energy exchange with the rest of the Universe...

- "Microcanonical" (purely Newtonian) dynamics conserves total energy...
- Langevin dynamics emulates both the intrinsic "viscosity" of the environment (friction term slowing down fast atoms, controlled by  $\gamma$ ) and occasional energy gains due to stochastic collisions with environment atoms (controlled by  $\eta$ )

$$\mathbf{F}_{i} = m_{i} \frac{\mathrm{d}^{2} \mathbf{x}_{i}}{\mathrm{d}t^{2}} = -\nabla E(\mathbf{x}_{i}) - \gamma m_{i} \frac{\mathrm{d}\mathbf{x}_{i}}{\mathrm{d}t} + \eta \sqrt{m_{i} T_{0}} \delta, \ \delta \sim \mathcal{N}(0, 1)$$

 Berendsen's empirical "coupling to a thermal bath" basically does the same, with a different formalism.

$$\mathbf{F}_{i} = m_{i} \frac{\mathrm{d}^{2} \mathbf{x}_{i}}{\mathrm{d}t^{2}} = -\nabla E(\mathbf{x}_{i}) - \frac{1}{\tau} \left(1 - \frac{T_{0}}{T}\right) m_{i} \frac{\mathrm{d}\mathbf{x}_{i}}{\mathrm{d}t}$$

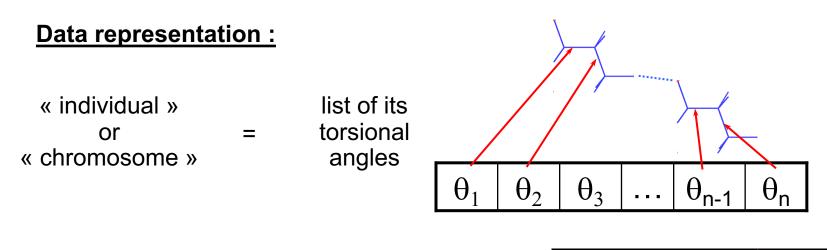
 Nose-Hoover's formalism is formally proven to be correct from a statistical physics standpoint... but quite tricky to use! Now, a molecule is not interested in *effectively* exploring its conformational space...

- So, it must be pushed towards more efficacious sampling, by creatively "tampering" with Newton's equations...
  - <u>Tabu or Memory dynamics</u>, a.k.a "Poling": forces the trajectory to become a self-avoiding walk, by means of bias potential terms that are large and positive for geometries similar to already visited ones, and zero outside the already explored phase space domain!
  - Potential Smoothing approaches: Artificially lower the high energy barriers of the energy landscape, without modifying the low energy zones – thus ensure for faster transitions...
  - <u>Adaptive bias</u>: add an extraneous force coupled to some global geometric descriptor (say: gyration radius), thus compelling the algorithm to generate both unfolded, linear and folded, globular geometries
  - <u>Umbrella sampling</u>: force the MD trajectory to follow a predefined "reaction path"... if you have a clue on how to define it!



# **Genetic Algorithms**

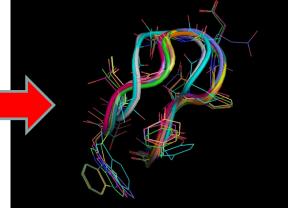
- Applying a Darwinian Evolution Scenario to a population of vectors ("chromosomes") encoding the solution to a problem
- Solution Quality is the "Fitness" score, and the fittest survive...



#### Population of individuals :

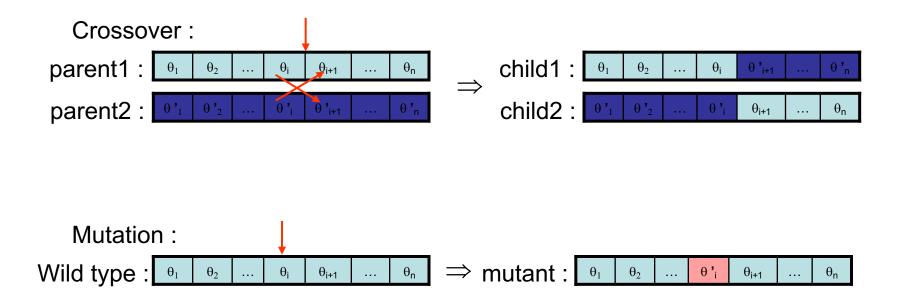
...

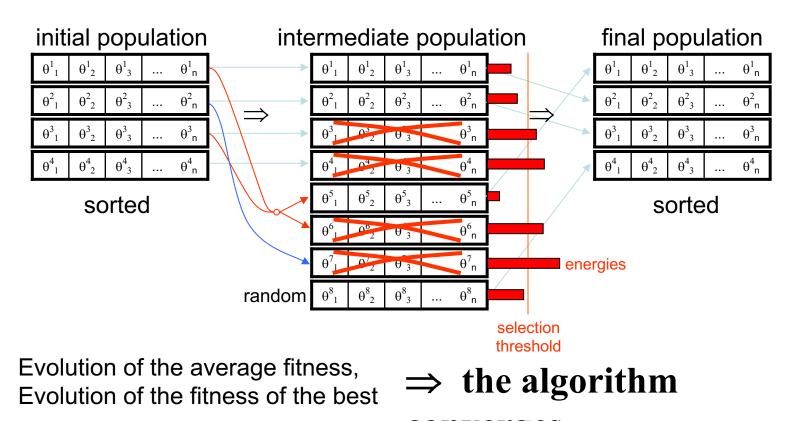
$\theta^1_{\ 1}$	$\theta_2^1$	$\theta_{3}^{1}$	 	 	 	 $\theta^1_{\ n}$
$\theta_1^2$	$\theta_2^2$	$\theta_{3}^{2}$	 	 	 	 $\theta^2_{\ n}$
$\theta_1^3$	$\theta_2^3$	$\theta_{3}^{3}$	 	 	 	 $\theta^3_{\ n}$
$\theta^4_{\ 1}$	$\theta_{2}^{4}$	$\theta_{3}^{4}$	 	 	 	 $\theta^4_{\ n}$





#### **Generation of new offspring :**



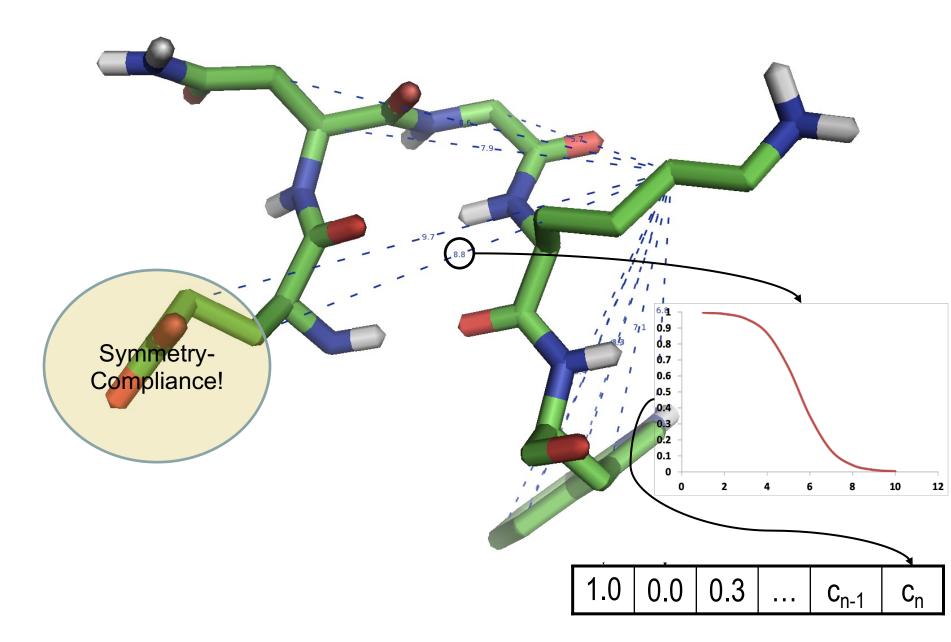


# Population Diversity Control is a Key Issue

38

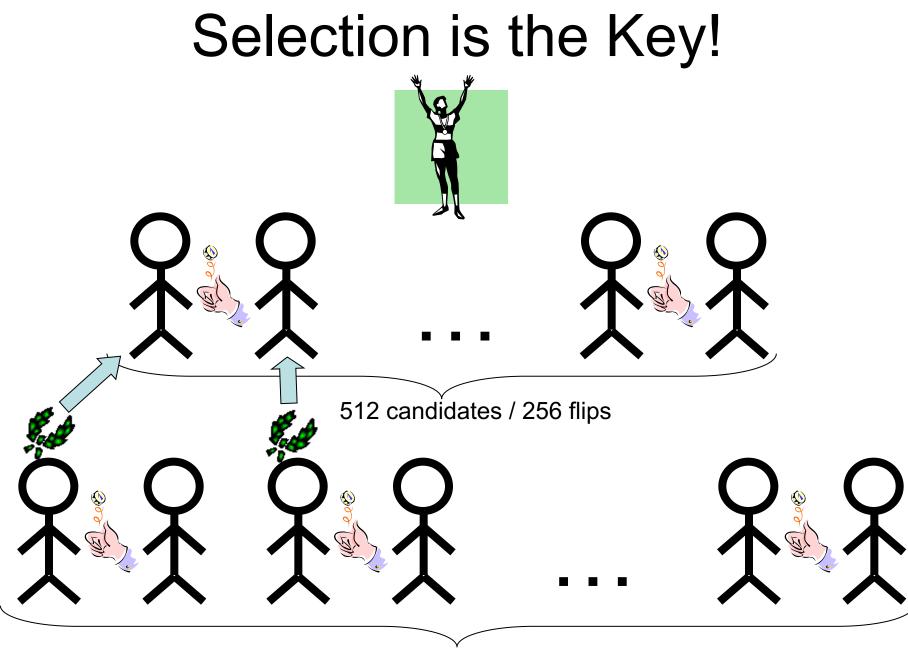
- Multiple 'Island' models parallel simulations occasionally swapping solutions
- Discarding of redundant chromosomes (requires a metric defining how similar two encoded solutions are!)

## Interaction Fingerprints to check for redundancy



Genetic Algorithms: Chance, Selection & the Coin Flipper's bet!

- Any problem admitting a vector as a solution may be coded by a "chromosome" and left in the hands of Darwin...or God??
- I bet (1M€) I can find a person who won a coinflipping challenge 10 times in a row, at his/her first attempt!!
  - In order to fulfill my promise, I need a total of 1024 coin flips to happen,
    - 1024/10=102 pretendents, each with a chance of (1/2)<sup>10</sup> to score 10 successive winning coin flips: ~90% chance to loose 1M€!
    - If you read "Darwin's Dangerous Idea" by D.C.Dennett, you are not allowed to bet !!



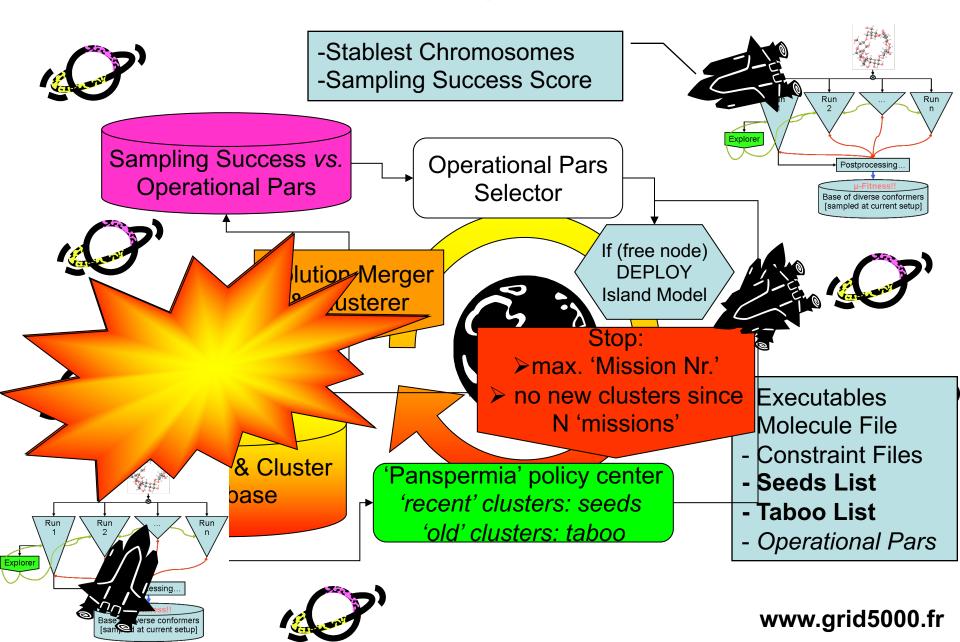
1024 candidates / 512 flips

Search for Optimal Sampling Setups in the Strategy Parameter Space...

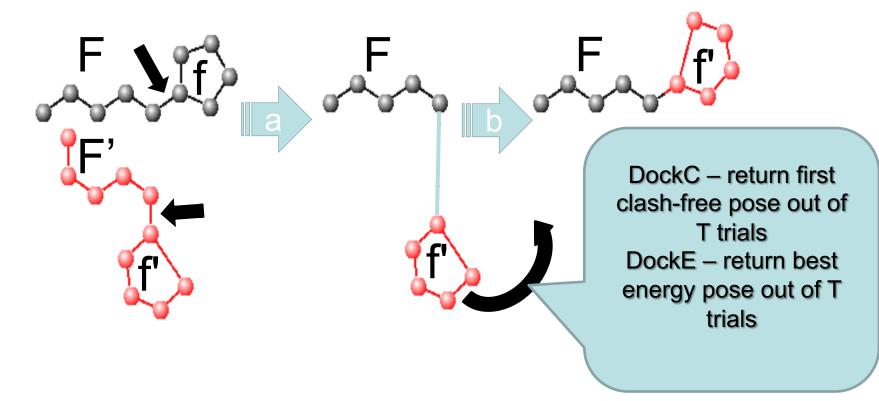
Population management													
Evolution management													
H		Selection pressure											
L		Convergence management											
	_[]	Apocalypse (population reset) frequency											
		Elitism											
		Global stop condition											
		$p_1$ $p_2$ $p_3$ $p_4$ $p_5$ $p_6$	p <sub>14</sub>	р <sub>15</sub>									

$$SamplingSuccessScore = f(p_1, p_2, ..., p_n)$$

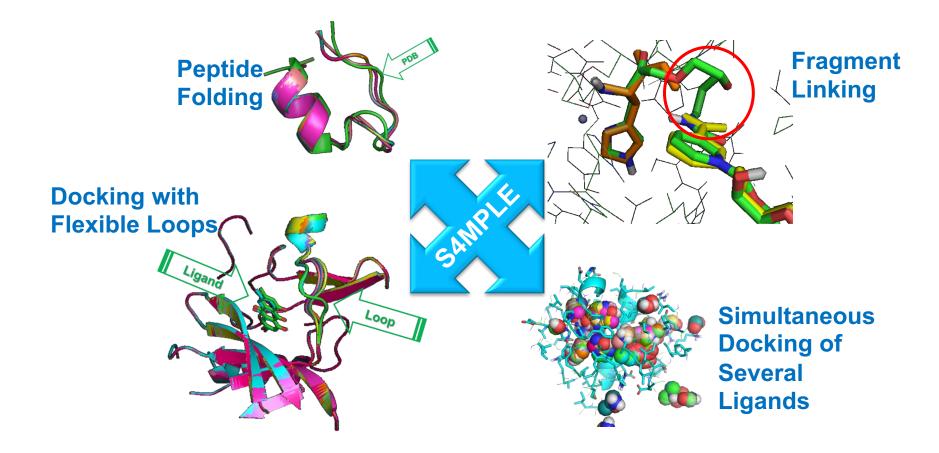
# GRID 5000-based 'Planetary' Model



Essential: you do not NEED to stick to torsions: Genetic Operators in Fully Flexible Mode (S4MPLE)



 If fragments are not bound, a putative favorable contact is used instead of the bond! Evolutionary Computing allows a unified approach to conformational sampling...



- Presentation Outline
  - The Basics: Molecules have Geometries!
    - Intramolecular energy: the Empirical Force Field
  - Sampling Methods: a brief overview
    - Molecular Dynamics: Walking like a molecule
    - Monte Carlo: Molecular Casino
    - Evolutionary Methods: in God/Darwin we trust!
  - Conclusions

# Conclusions

- Conformational Sampling is the Key Element for Understanding of Molecular Behavior
- It may range from very simple to extremely difficult, to impossible
- If you don't do it well, better don't do it at all: empirical methods based on molecular topology only may be more accurate than 3D models based on wrong – or too few – conformations
- Two main sources of errors: A.) wrong calculated energy-geometry landscape (poor Force Field parameterization) and B.) – insufficient sampling!
- Docking is just a specific case of conformational sampling, involving at least two molecules: a binding "site" and one or more "ligands"
- You will often hear that the knowledge of the "bioactive" conformer is paramount to understand binding. This is necessary, but sometimes not sufficient.
- Entropic effects perversely insist on being important, in spite our inability to properly estimate them!