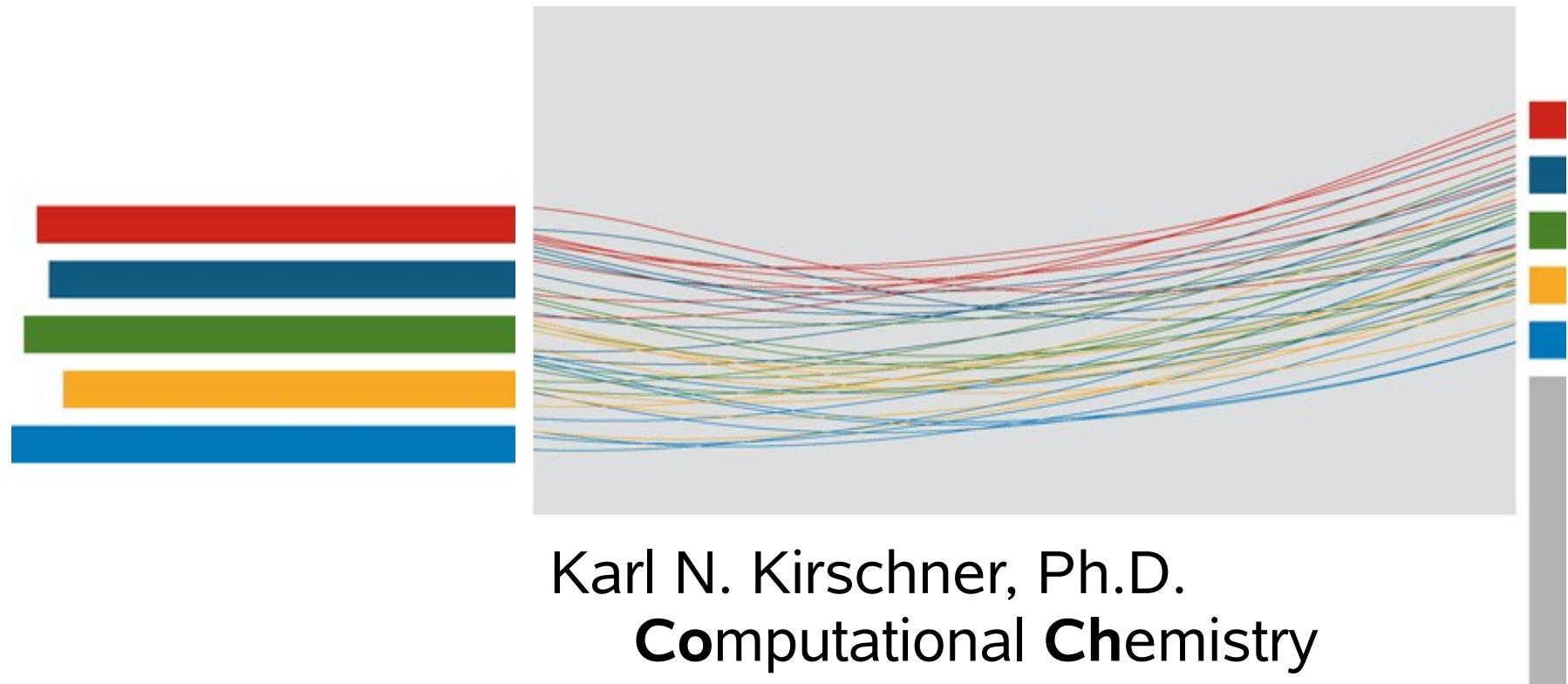


Connecting Quantum Mechanics to Coarse-Grained Simulations: the Role of Force Field Development



Synopsis of SCAI and the “CoChE” Group

Part of “Instituzentrum Birlinghoven” -
One of the largest computer science
research centers in Germany

SCAI: ~90 Employees + 40 PhD
Students; highly interdisciplinary

2008 Annual budget : 8.6 Mio. Euro

40% of the SCAI budget from
industry

CoChE founded in 2007: 7 senior staff
(PhDs), interdisciplinary 2 PhD-
Students and 3-5 Undergraduates



Universität zu Köln – Lehrstuhl Angewandte
Mathematik / Wissenschaftliches Rechnen



Bonn Aachen International Center
for Information Technology
Life Science Informatik (Prof. Hofmann-Apitius)



DLR-SISTEC – Simulations- und Softwaretechnik
(Leitung: Prof. U. Trottenberg, R. Hempel)

Verteilte Systeme und Komponentensoftware
Software QS und eingebettete Systeme



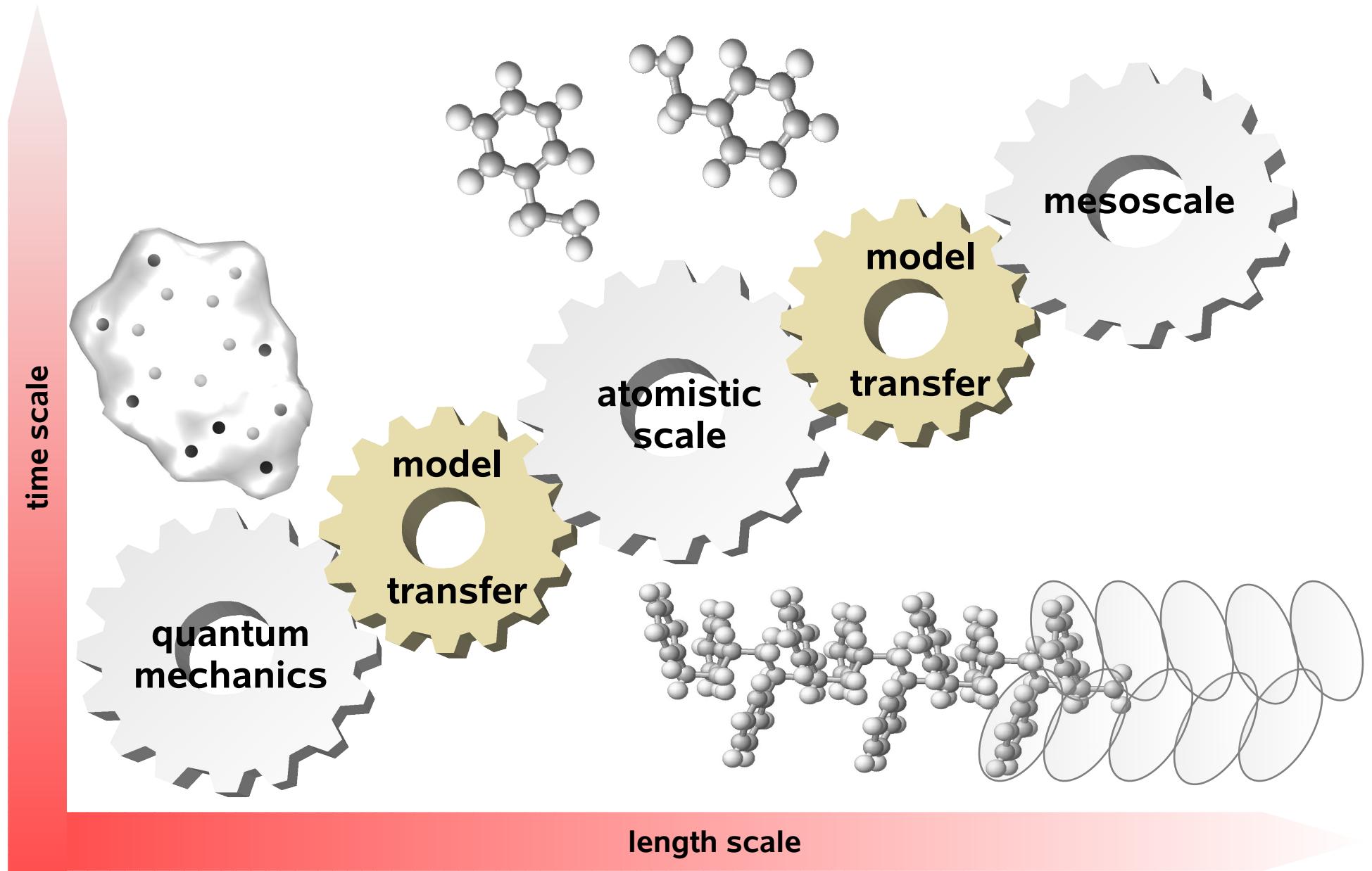
Force Fields: an old topic, but remains important

Different routes / methods / philosophies exists for their development --- “The art of parameterization”

Challenge: From our experiences and the data in the literature, can we begin to formulate more standard routes towards developing and optimizing a force field in a timely and reliable fashion?

Why are we thinking about this? Industry applications & contribute to the scientific community.

Soft matter - Multi-scaling modeling

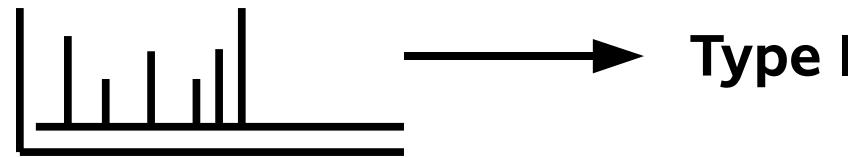


What is the Current State?

Basic atomistic force field equations appear to be established

$$V = \sum_{\text{bonds}} K_r (r - r_{eq})^2 + \sum_{\text{angles}} K_\theta (\theta - \theta_{eq})^2 + \sum_{\text{dihedrals}} \frac{V_n}{2} [1 + \cos(n\phi - \delta_n)] + \sum_{i,j} \left[\frac{A_{ij}}{r_{ij}^{12}} - \frac{B_{ij}}{r_{ij}^6} + \frac{q_i q_j}{\epsilon r_{ij}} \right]$$

Types:



Molecular Classes:

- Functional/Small molecule (transferable) → Hydrocarbon, Alcohol, Esters...
- Specialized (nontransferable) → Solvent, Molecule specific
- Biological (something in between??) → Protein, DNA, Carbo.

Current research towards how to including additional quantum effects
(eg polarization)

Different routes exist towards developing new parameters

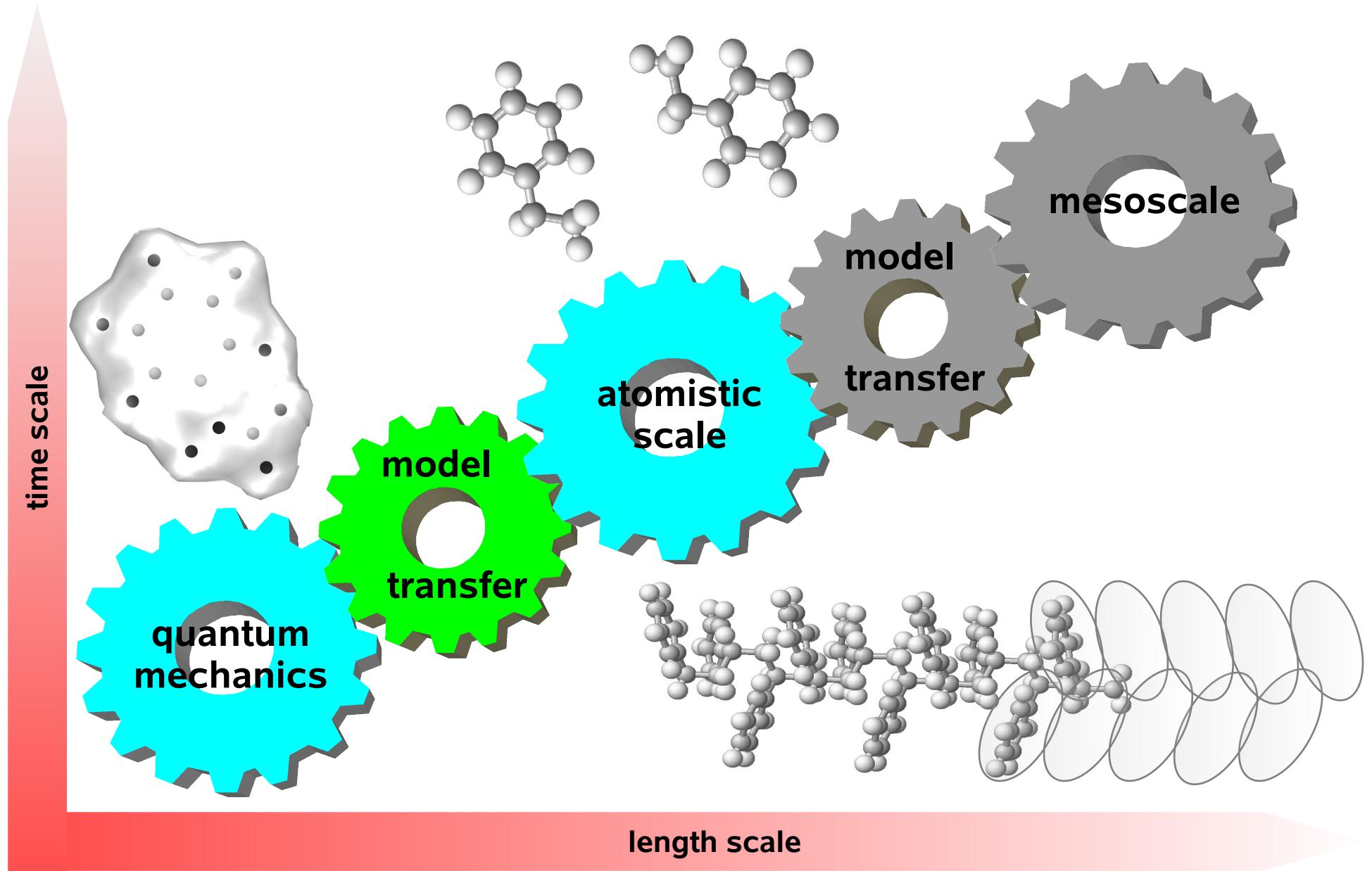
- Fully manual method
- Cross terms
- L.-J. Parameters
- QM – which levels
- Keep parameters close to chemical intuition vs completely adjustable
- Weighting specific molecular conformations
- How is experimental data used? a) training b) validation
- What exp observables are used -geom, energy, ΔH_{vap}°
- Generic or Explicit torsion terms (eg $X-C_{sp^3}-C_{sp^3}-X$ vs $C_{sp^3}-C_{sp^3}-C_{sp^3}-C_{sp^3}$)

Different programs use different force fields (eg MM3, Parm99SB, GROMOS, and ports)

Internal default settings effect the force field performance (eg scaling factors)

Challenge: Develop a time-efficient and systematic route for new reliable parameters.

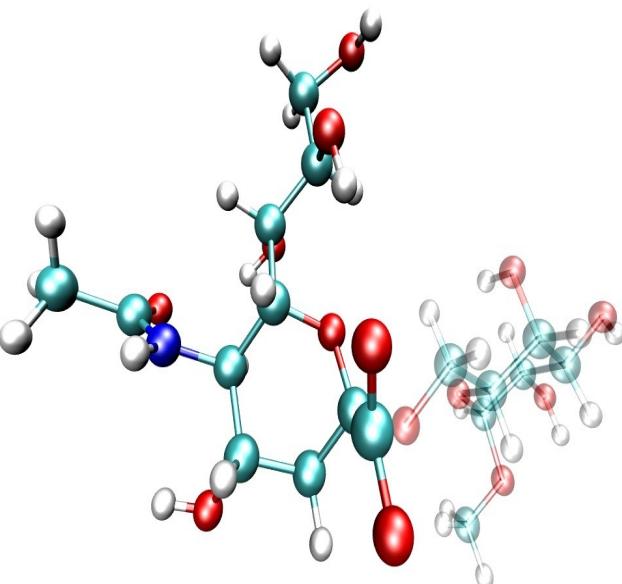
Soft matter - Multi-scaling modelling



**Current and future Goals via the lessons learned so far (atomistic)**

- Have a *Development* and *Validation* suite of molecules
- Isolated chemical families for parameterization
- Use a basic form of the FF Equation, *initially* (eg no cross terms)
- Keep the number of atom types as low as possible
- Parameters should be transferable & **program independent**
- Remove 1-4 nonbonded & electrostatic scaling factors
- Use QM to develop the force field parameters (at least for version 1)
 - HF/6-31G(d)//HF/6-31G(d) → *Geometries*
 - MP2/ or B3LYP/6-31++G(2d,2p)//HF/6-31G(d) → *Energies*
 - B3LYP/ or HF/aug-cc-pVTZ//HF/6-31G(d) → *Electrostatics*
- Partial Atomic Charge Sets → **Ensemble averaged** or Boltzmann weighted (via **RESP**)
- Explicit parameters (C_{sp^3} - C_{sp^3} - C_{sp^3} - C_{sp^3} vs $X-C_{sp^3}-C_{sp^3}-X$)
- Final test of parameters is the comparison between theoretical and experimental results

Provide a firm foundation for a) future **parameter refinement**,
b) adding e- lone pairs, and c) adding polarizability



Chemical Functional Groups:

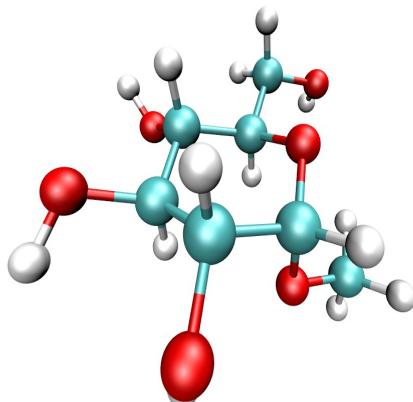
Hydrocarbons	Ether Amides
Alcohols	Alcohol Amides
Ethers	Ether Alcohols
Amides	Ether Carboxylates
Carboxylates	

Develop the parameters in a systematic step-wise fashion → allows the physics to be isolated to each parameter

1-4 Scaling Factors (AMBER)

Why scaling factors:

it was justified “given the known overestimation of short-range repulsion by a 6-12 form of potential.”†



Why it is a problem:

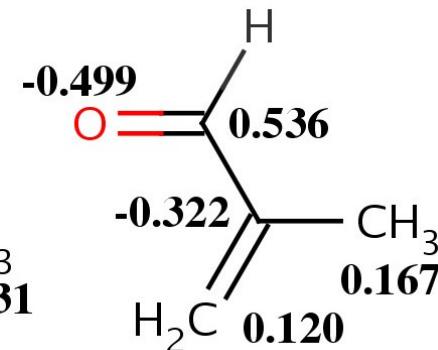
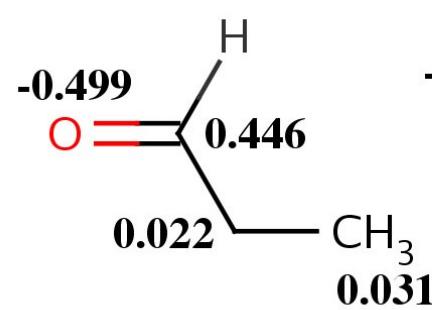
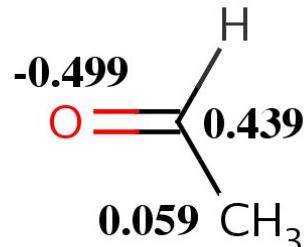
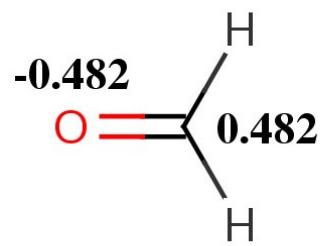
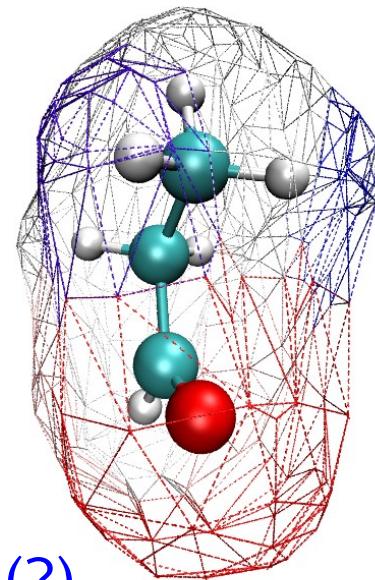
1-5 interaction ≡ a 1-4 interaction ($\text{O}\cdots\text{O}$ equal dist)
captured by torsions parameters

† Cornell et al. *J. Am. Chem. Soc.* 115 (1993) 9631 & 115 (1995) 5179.

‡ Kirschner et al. *J. Comput. Chem.* 29 (2008) 662.

RESP Model†

- Input: QM ESP (CHELPG & large basis set)
- Uses a hyperbolic restraint (varied by a weighting factor)
→ purpose is to keep the P.A.C lower magnitude
- Helps to correct the high dependency of conformation in P.A.C. due to buried atoms
- Consistent P.A.C. in molecular *fragments*
→ allows for better generation of torsion parameters & L-J (?)

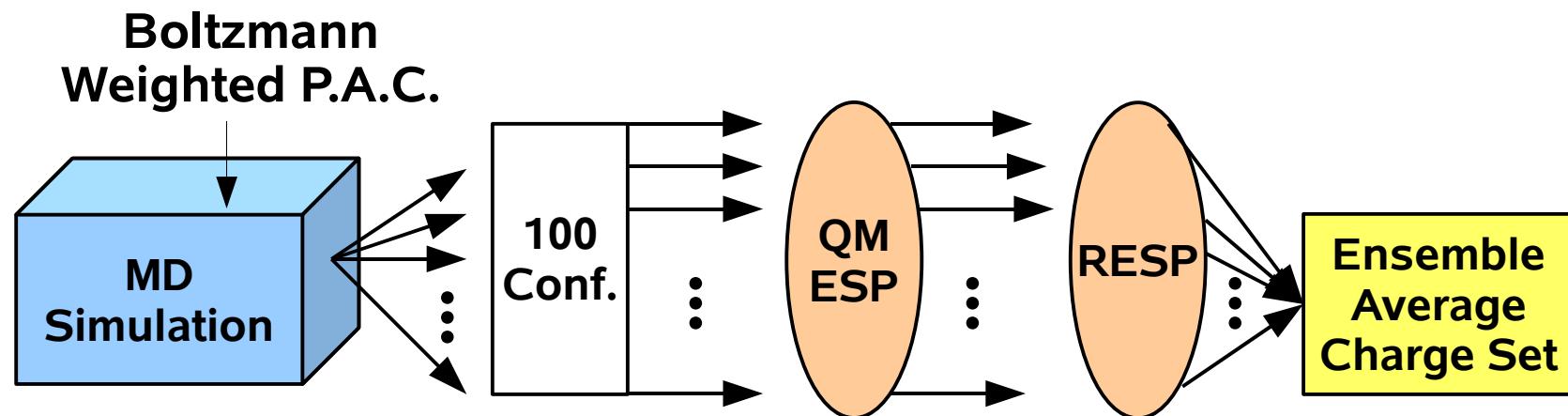


- Boltzmann Weighting vs Ensemble Averaging
 - Boltzmann Weighting: *gas-phase* QM relative energies
 - Ensemble Averaging: 1st approximationg to capturing *solution-phase* effects & conformations

Do not need every atom to obtain a good RESP fitting to the ESP
Charge are more chemically intuitive†

For carbohydrates, removal of aliphatic hydrogens result in optimal number for fitting‡ ... this holds true for other compounds as well

Methodology for final charge set (eg 100+ carbohydrate residues)

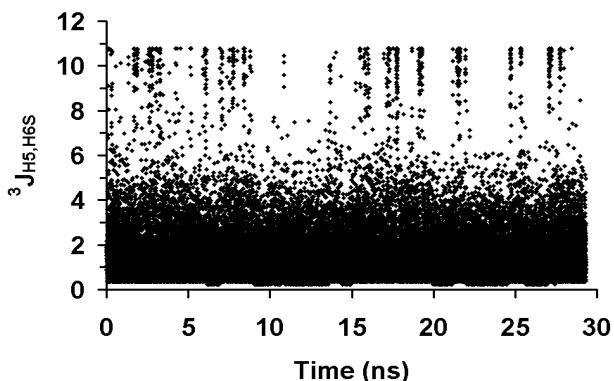
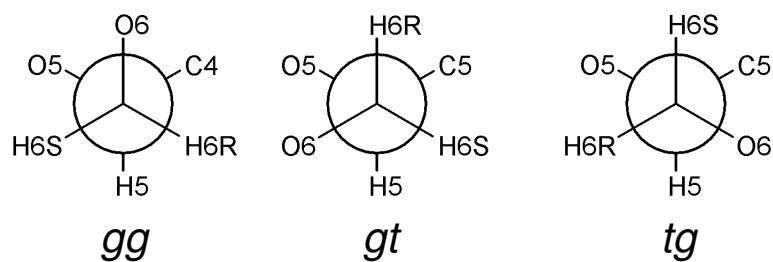
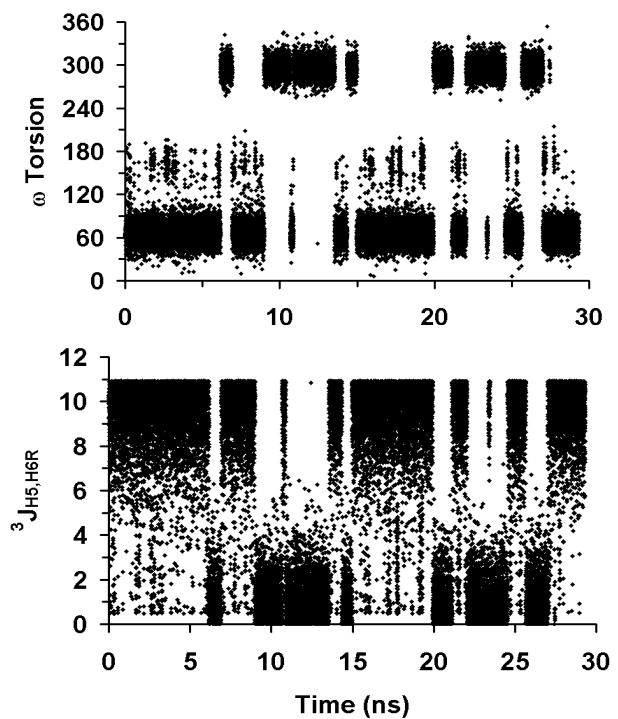
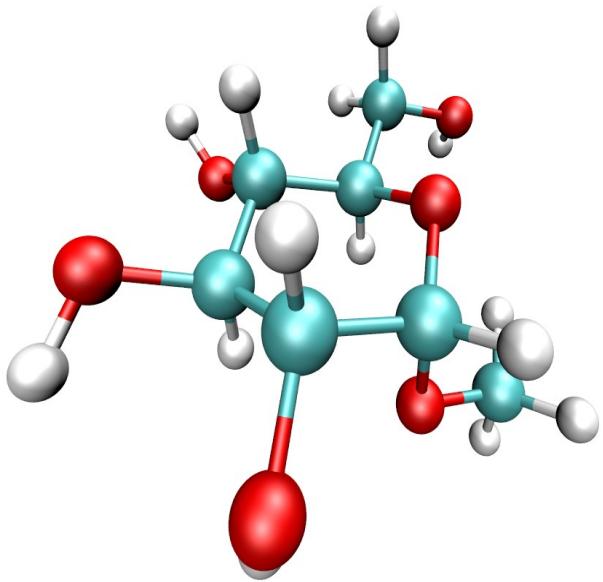


Take home point → a charge set that is less conformationally biased and reflects the solution phase

† Franci et al. *J. Comput. Chem.* 17 (1996) 367 & *Glycoconjugate J.* 14 (1997) 501.

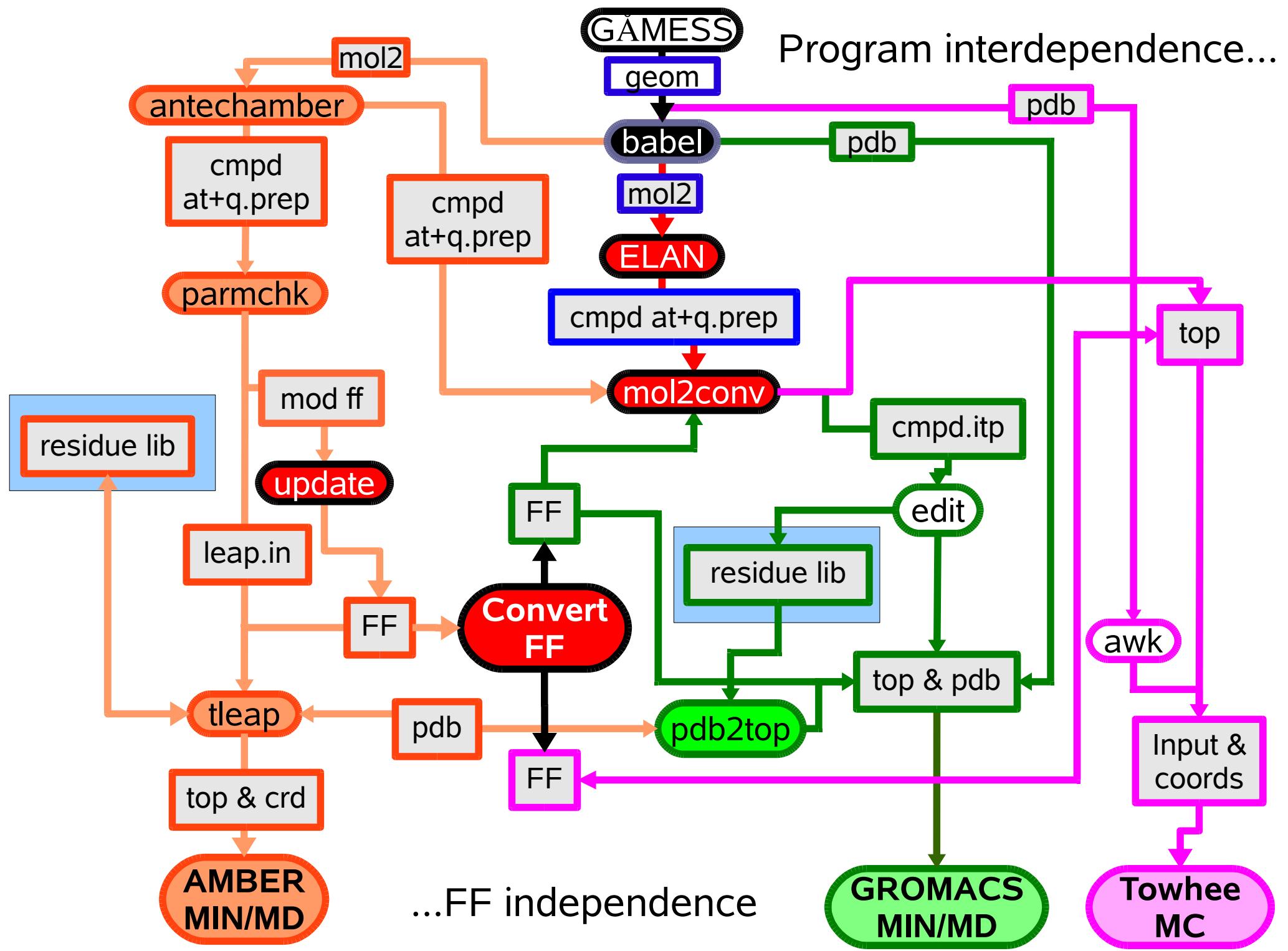
‡ Basma et al. *J. Comput. Chem.* 27 (2001) 1125.

‡ Kirschner et al. *J. Comput. Chem.* 29 (2008) 662.

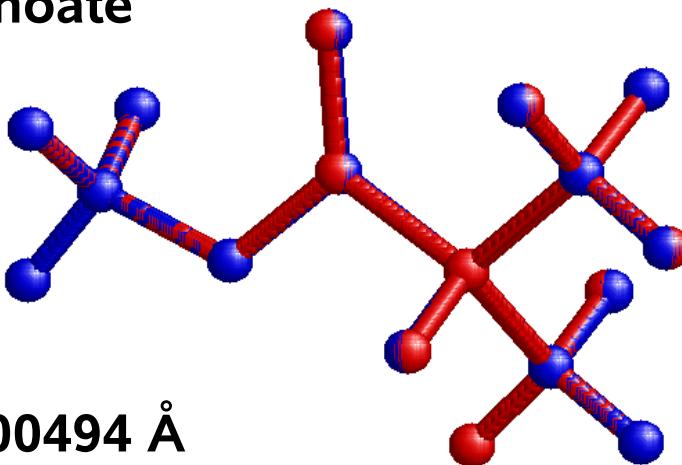


α-D-GlcPOMe	$^3J_{H5,H6R}$	$^3J_{H5,H6S}$
NMR (avg)	5.4	2.3
Glycam06	5.4 ± 1.7	2.9 ± 2.0
<i>gt</i>	9.5 ± 1.4	1.8 ± 1.2
<i>tg</i>	2.3 ± 1.6	9.0 ± 1.9
<i>gg</i>	1.0 ± 1.0	1.6 ± 1.1

Program interdependence...



Methyl 2-methylpropanoate

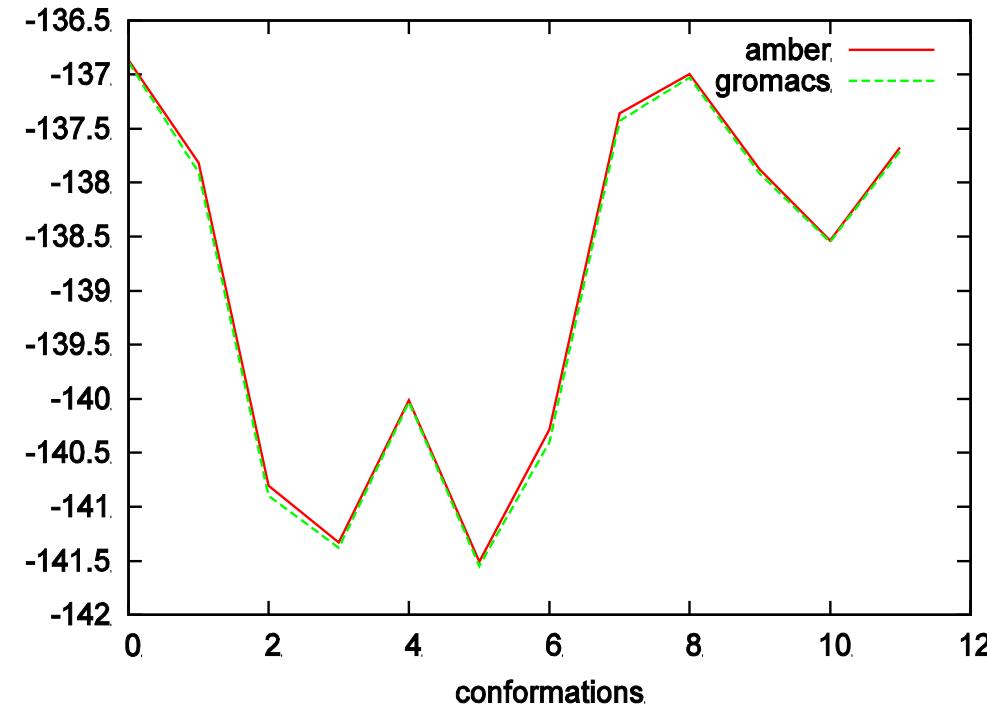
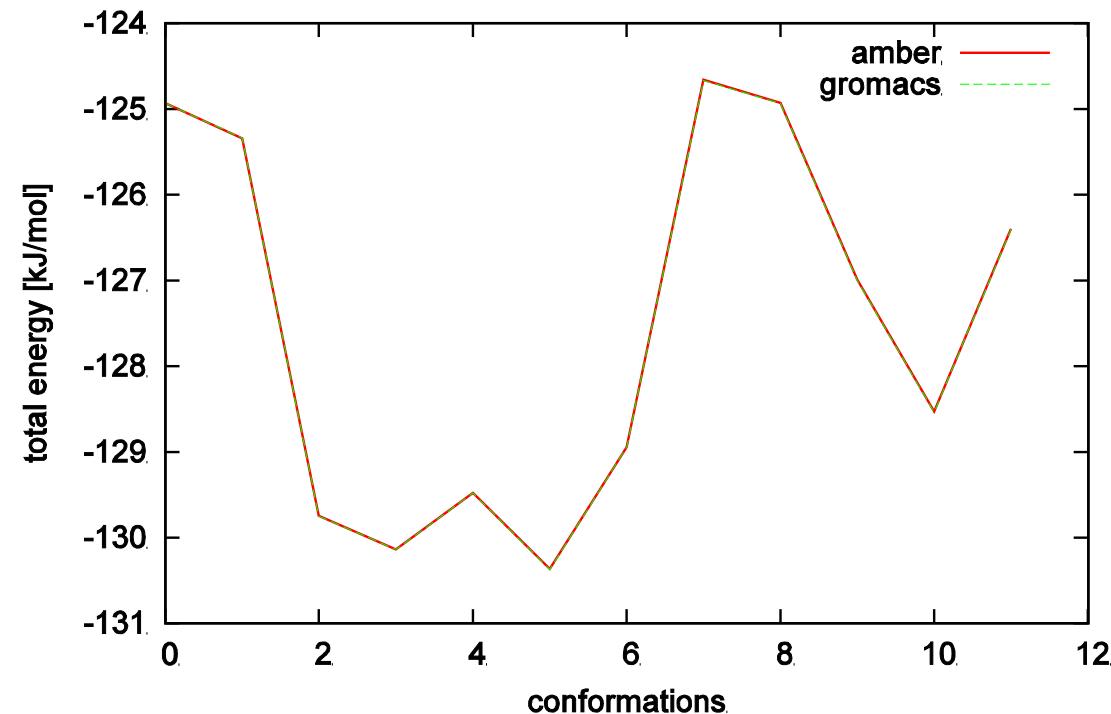


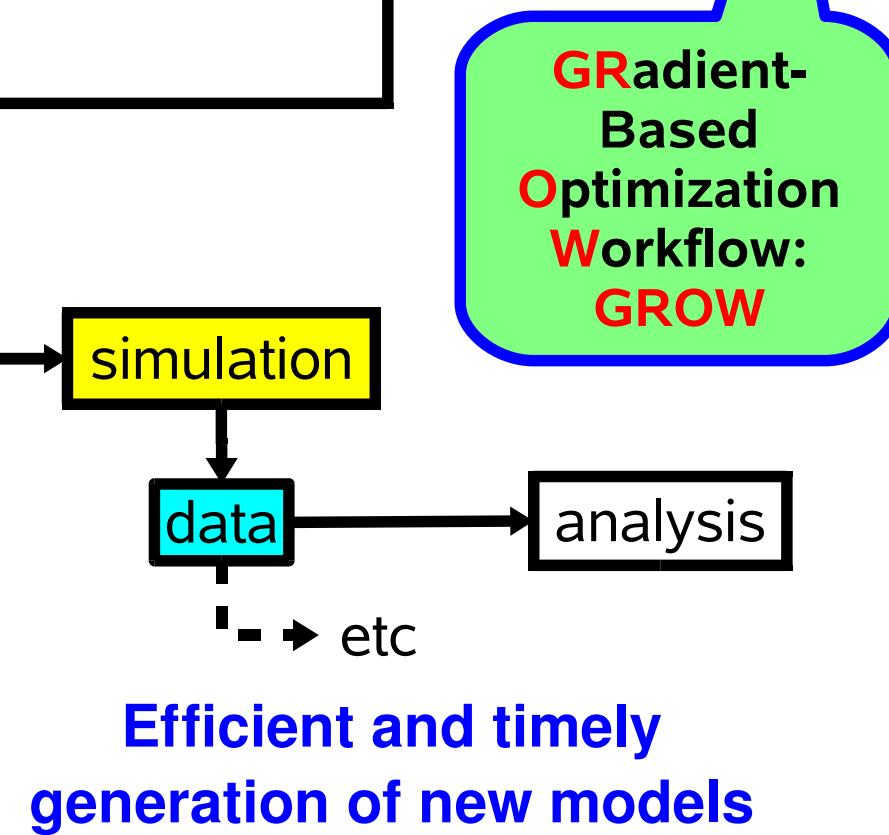
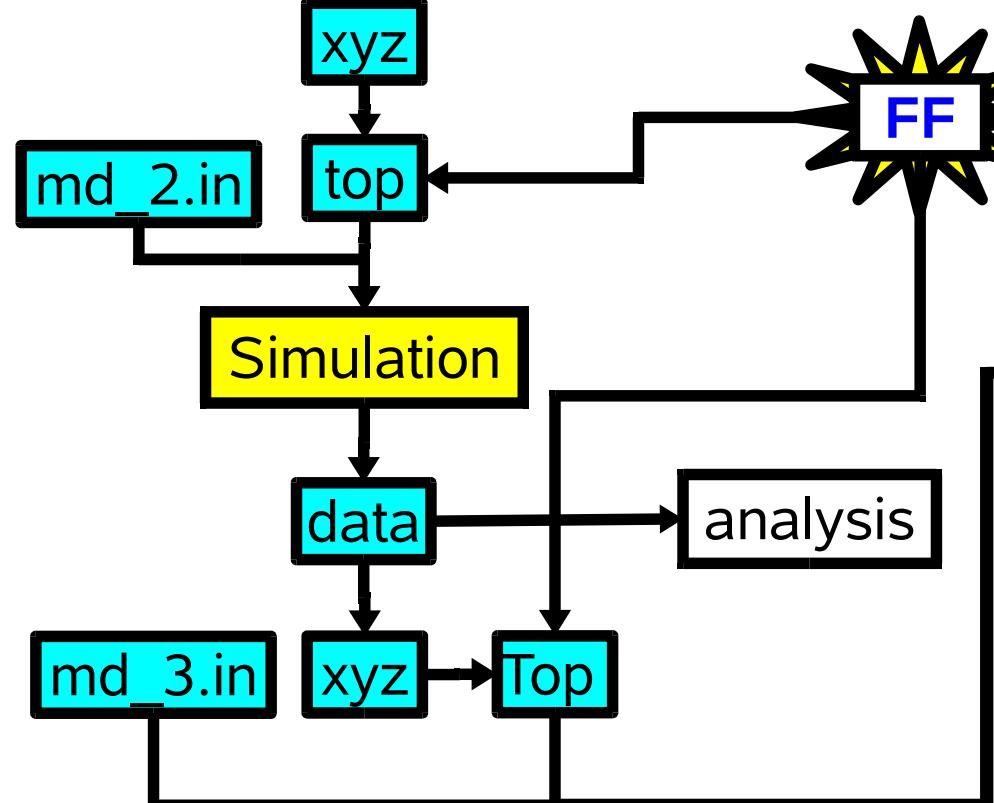
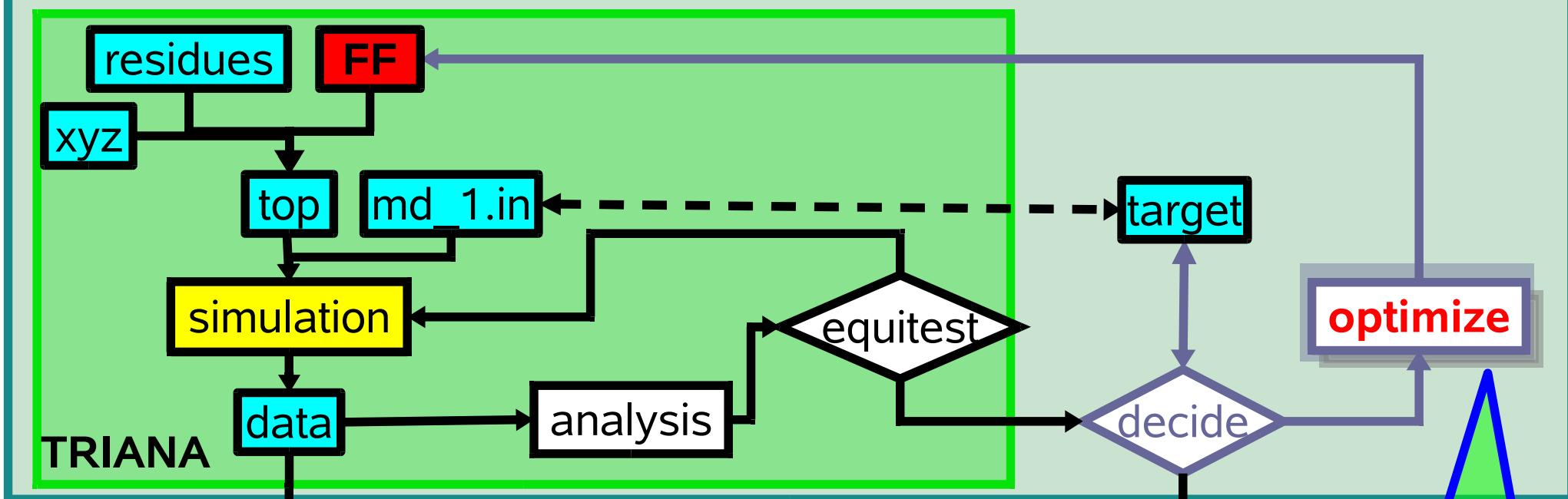
max. rms = 0.00494 Å

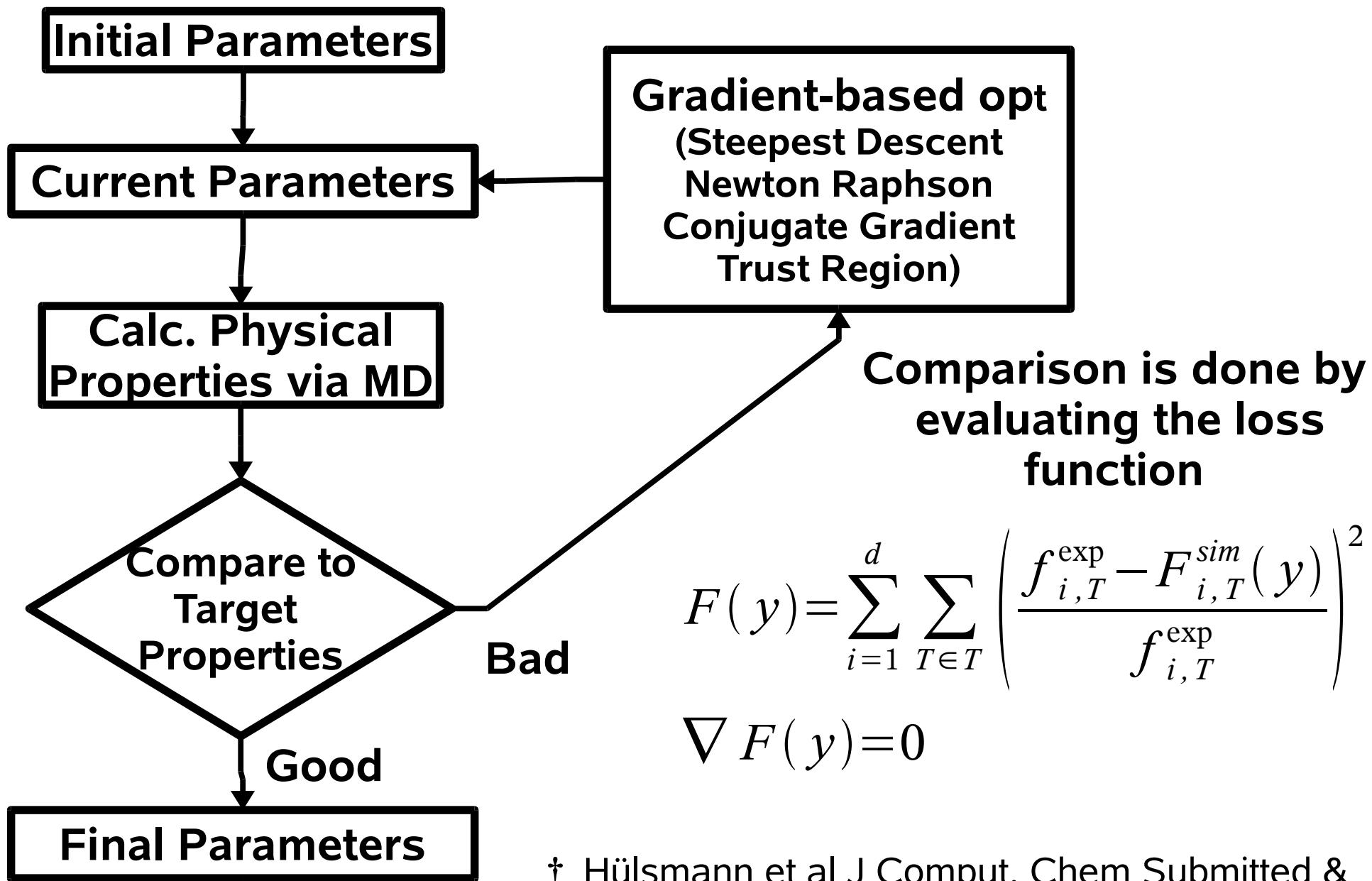
max. rms = 0.0196 Å

0 steps of minimization

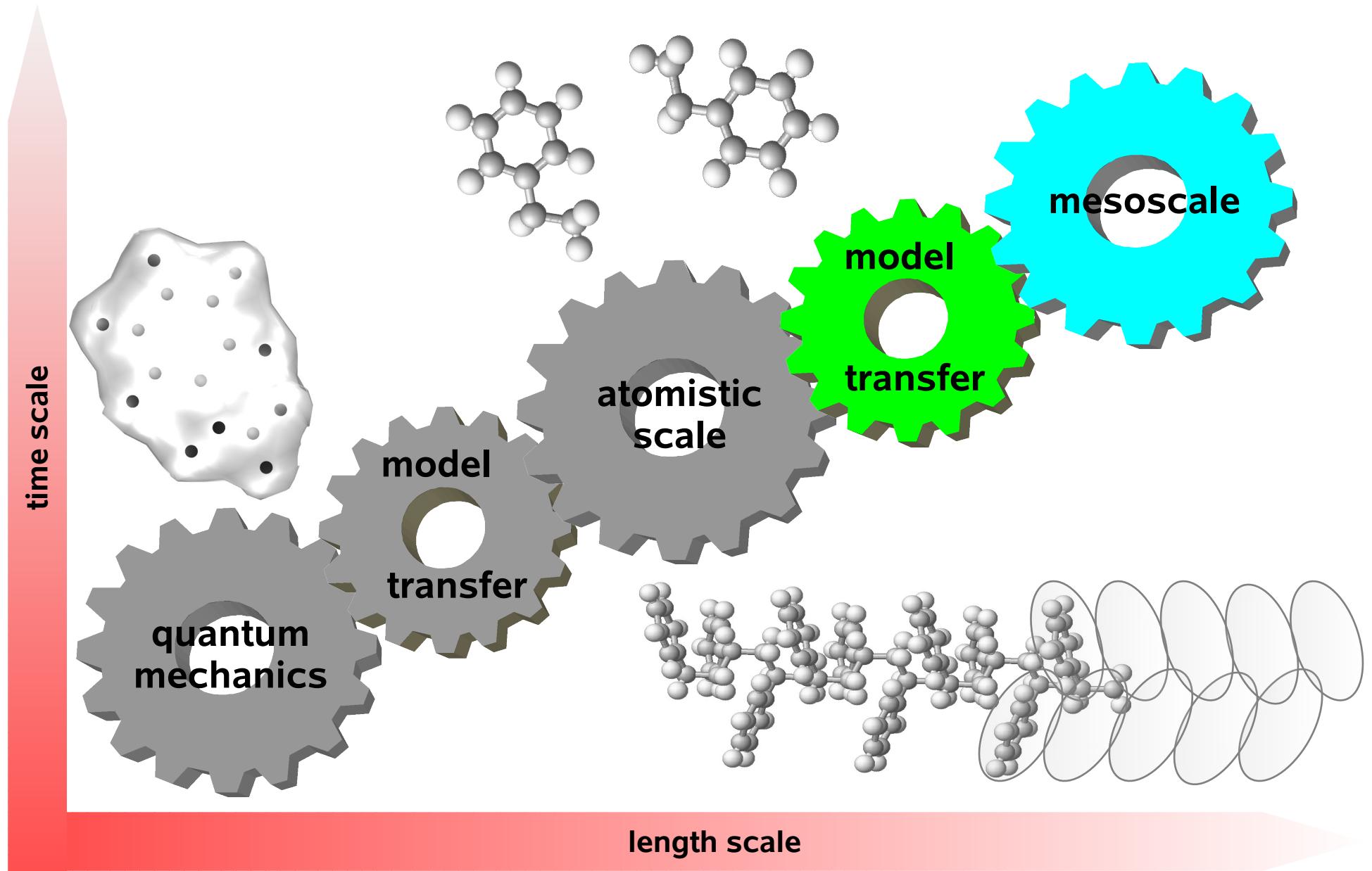
100000 steps of restrained minimization.

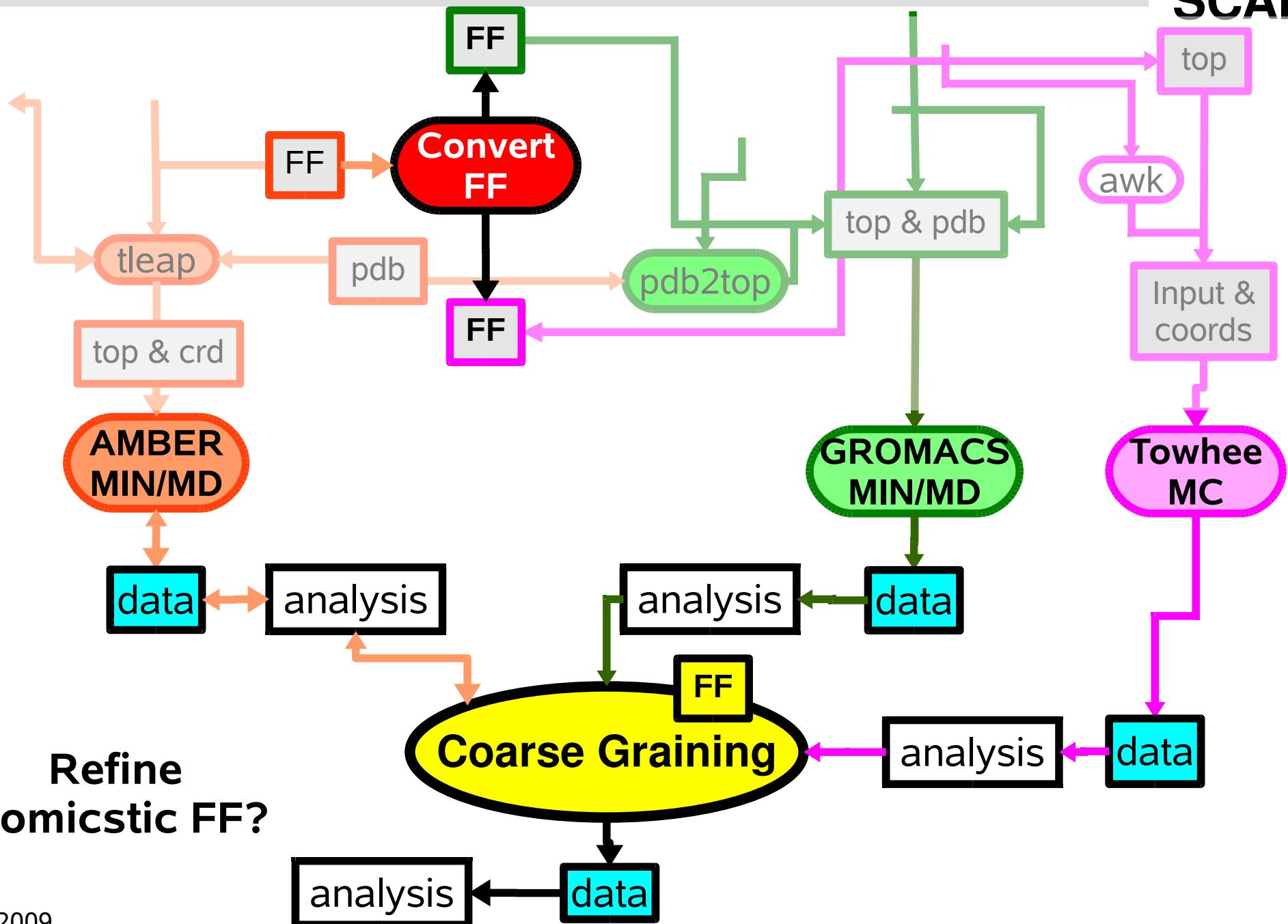






† Hülsmann et al J Comput. Chem Submitted &
Hülsmann et al Comp Phys Comm submitted





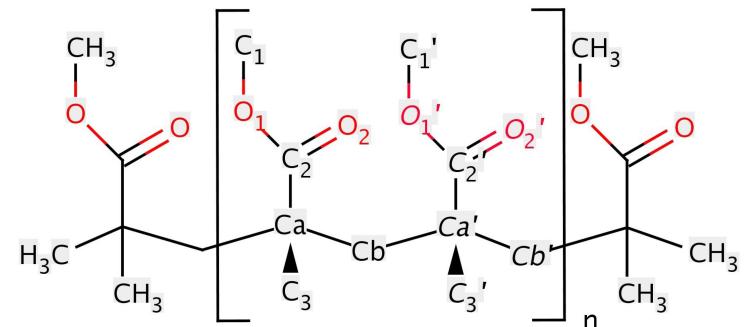
Atomistic model

Number of polymers: 59

Number of monomers per molecule: 8

Number of atoms: 7375

Temperature: 450 K

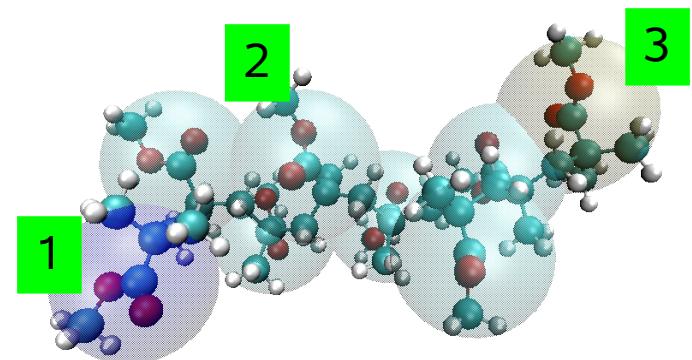


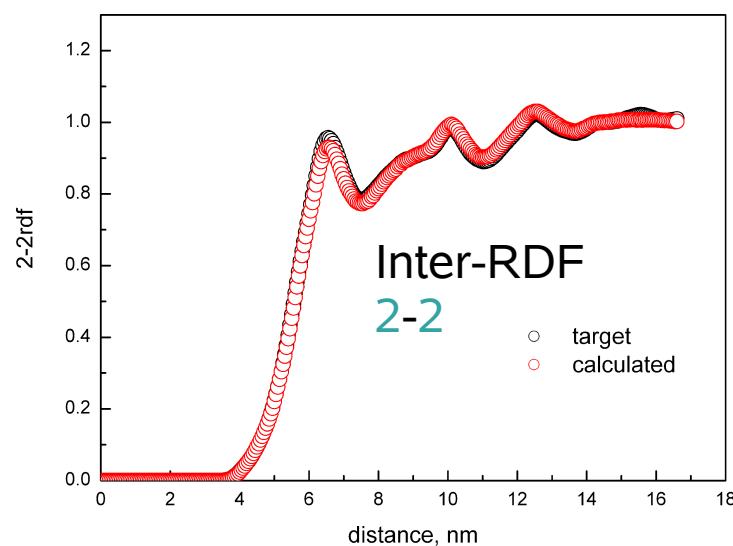
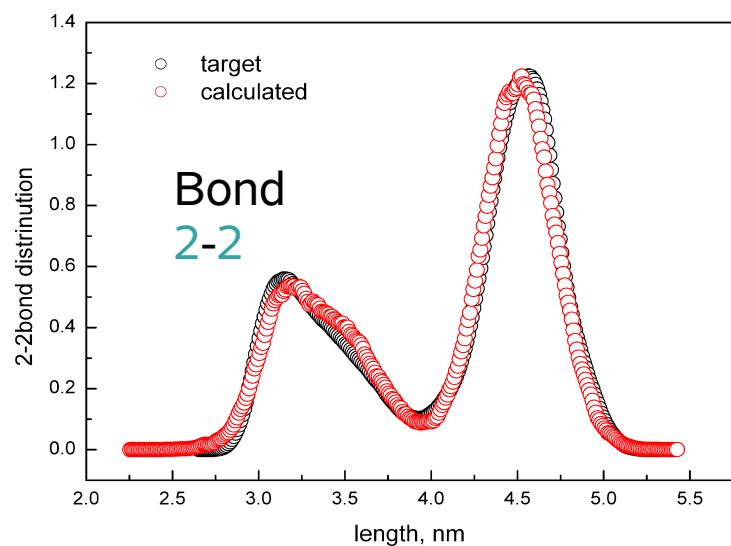
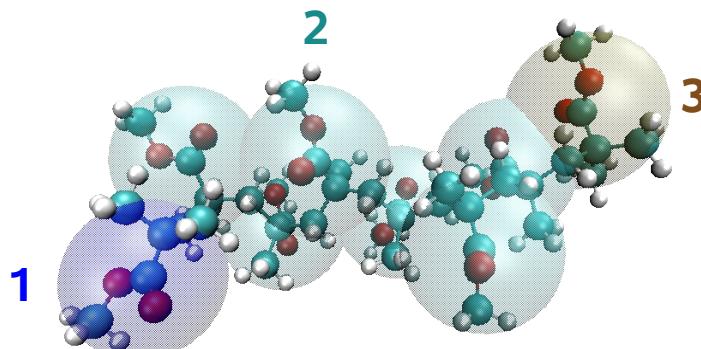
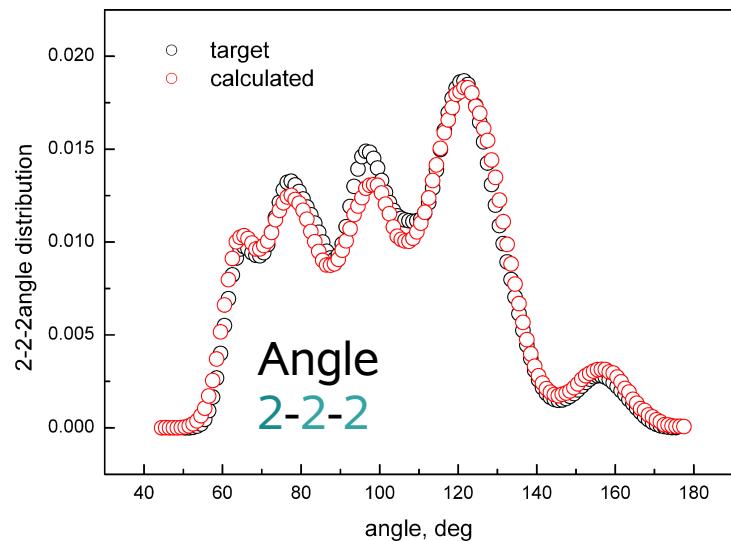
CG model

Number of beads: 472

Number of different type of the beads: 3

The C₂ position has been chosen
as center of the beads





We reproduce atomistic MD target distributions
for inter- and intramolecular interactions

ESPResSO: Extensible Simulation Package for Research on Soft matter,
MPI for polymer research (MPIP), Mainz

Open source, users in about 100 institutes all over the world

ESPResSo++: collaborative project of MPIP and Fraunhofer SCAI

Integration scheme:

- Standard MD (velocity-Verlet, constant energy)
- MD with constant T (eg Langevin) or constant P

Interactions

- Lennard-Jones, Morse, Gay-Berne (ellipsoids), tabulated potential
- Harmonic or nonlinear expander bonds, angle potentials, dihedrals

Spatial constraints: walls, spheres, cylinders

Electrostatics; Hydrodynamics via Lattice-Boltzmann; Analysis

FF Generation + Opt

- a) Isolate chemical functionalities
- b) Explicit parameters
- c) No nb or el 1-4 scaling
- d) RESP: Boltzmann, Ensemble Averaging
- e) In-house scripts

Hydrocarbon (un-saturated), Alcohols, Ester, Ethers, (Amides)



Program Independent

- a) AMBER, GROMACS, Towee, etc.
- b) In-house script converters for FF, residue lib., etc.
- c) Must test FF performance – min & MD

Coarse Graining

- a) Collaboration
- b) ESPResSo++

FF Refinement

- a) TRIANA Workflow to automate job cycle
- b) GROW: parameter optimization towards target goal



Acknowledgments & Coworkers



Glycam06

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Austin Yongye (UGA)

Coarse Graining

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Elena Algaer (TU Darmstadt)
Ali Karimi (TU Darmstadt)