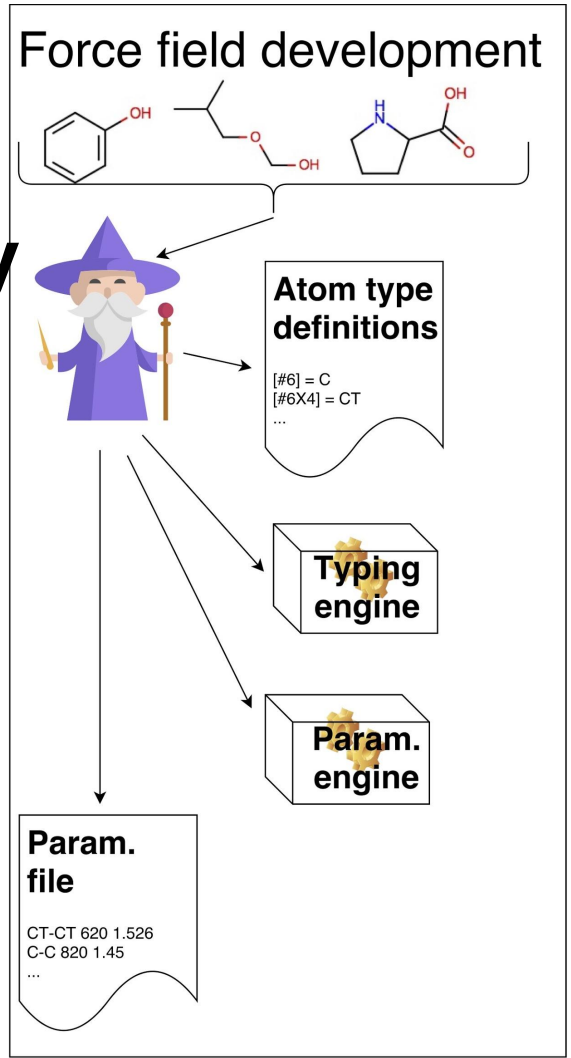
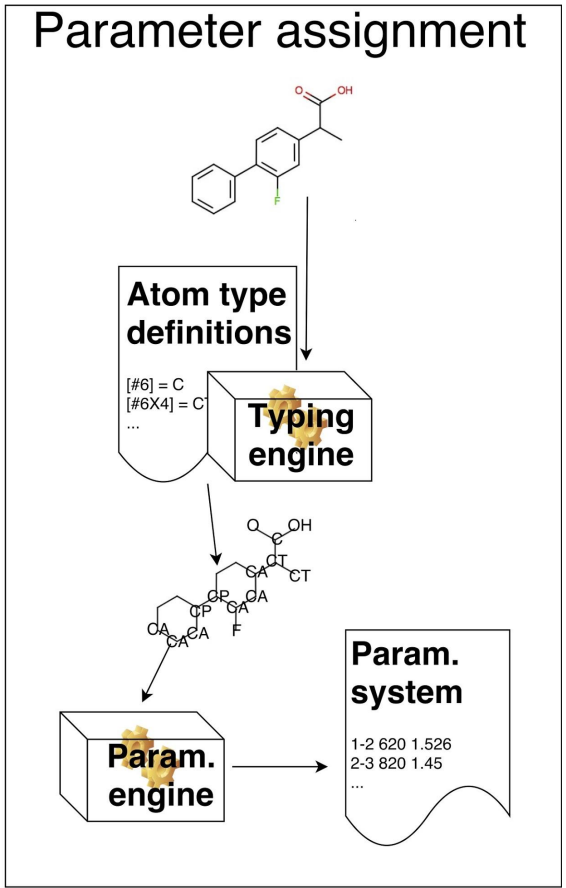
Chemical perception and SMIRNOFF typing

Caitlin C. Bannan and David L. Mobley (UC Irvine)

#smirnoff and **#smarty** on Slack

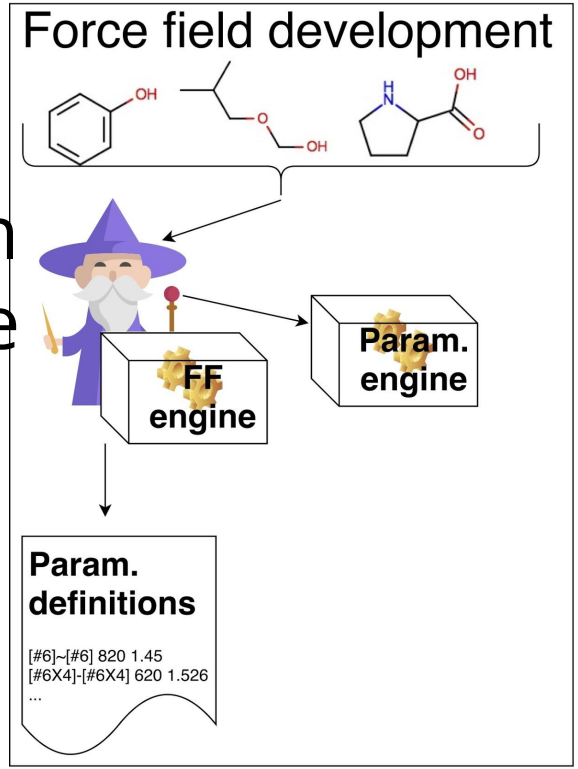
Force fields typically rely on “indirect chemical perception”, where atom typing is a design decision

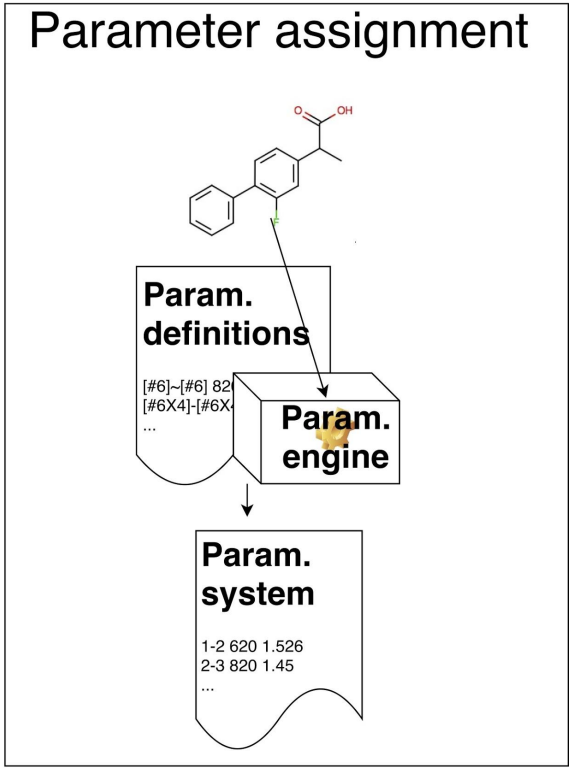




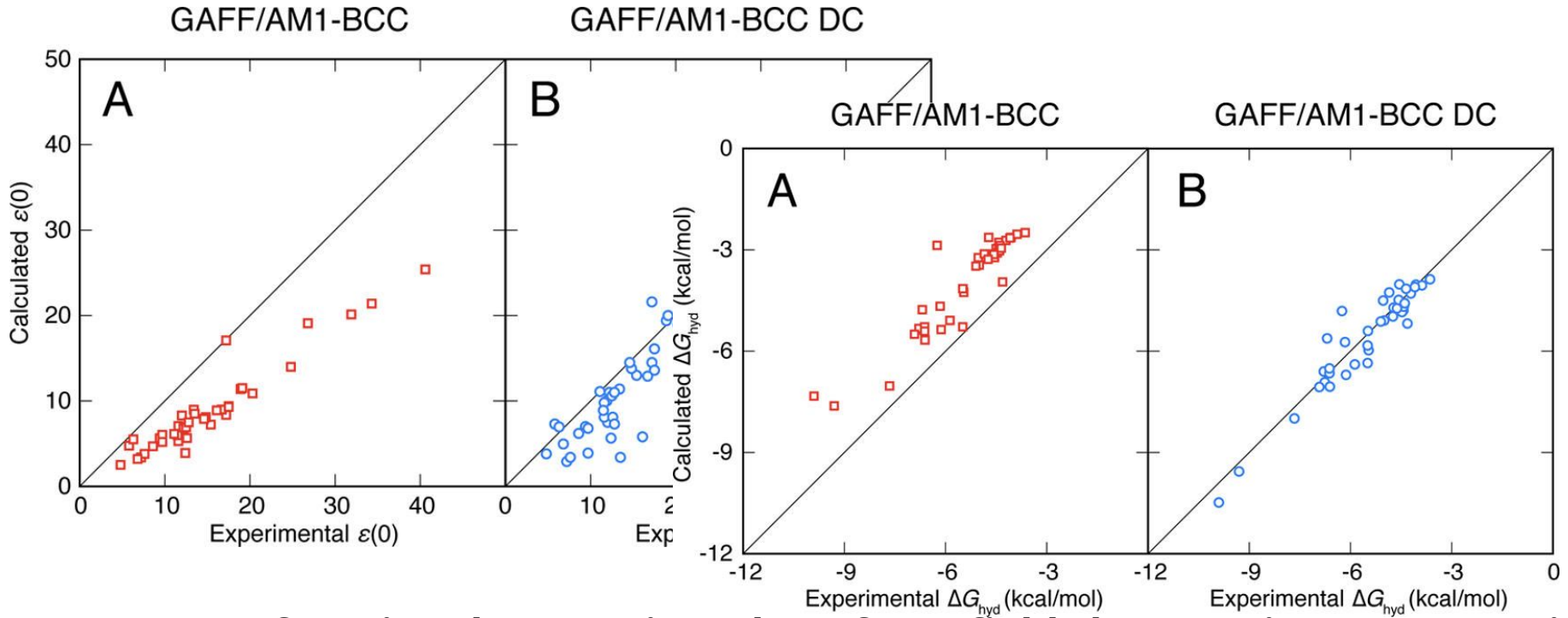
We employ direct chemical perception

The distinction is important: All chemical information is still available to the parameterization engine





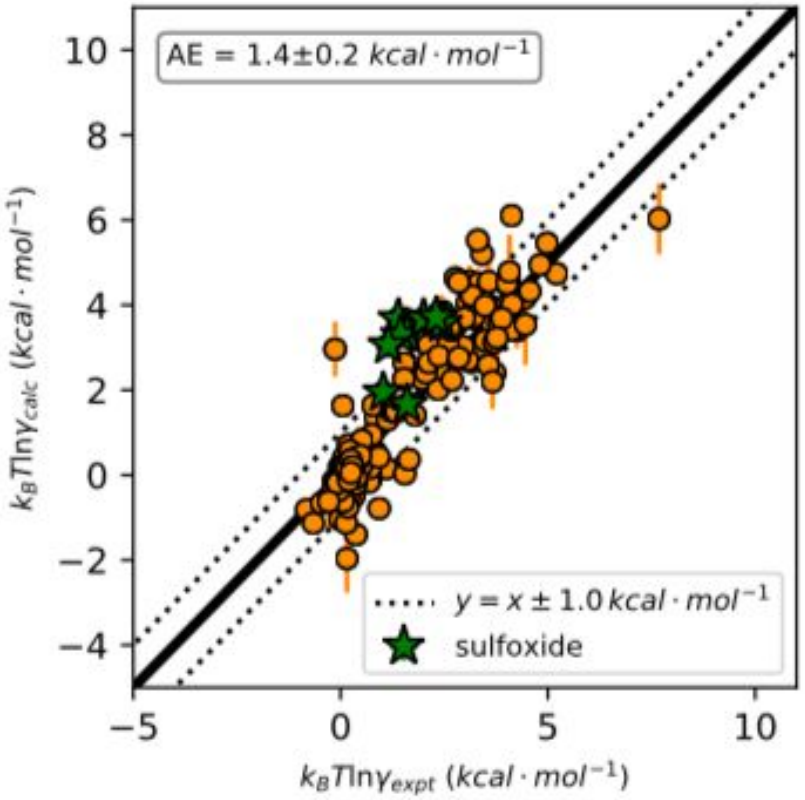
Why new force fields? We already know we can do better than today's force fields without new physics



Many functional groups in today's force fields have serious systematic errors that await a general fix, because refitting is too hard

DOI: 10.1021/jp411529h

Similar issues stare at us everywhere we look



Infinite dilution activity coefficients are untapped for force field development and show clear systematic errors

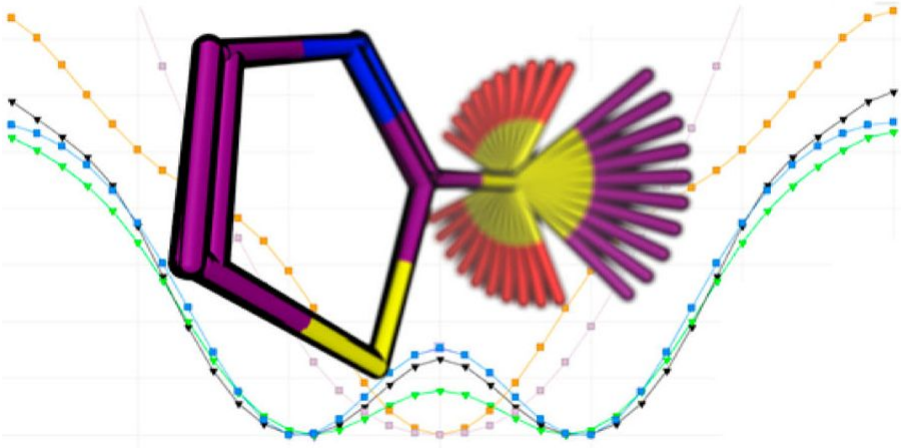
They inform on relative solvation in different solvents

Here, DMSO is poorly represented as a solvent

(Calculations of all the suitable activity coefficients from NIST's ThermoML done overnight on Orion)

https://doi.org/10.26434/chemrxiv.7199786.v2

We all have our own war stories, too



JOURNAL OF CHEMICAL INFORMATION AND MODELING

Article

pubs.acs.org/jcim

A Comparison of Quantum and Molecular Mechanical Methods to Estimate Strain Energy in Druglike Fragments

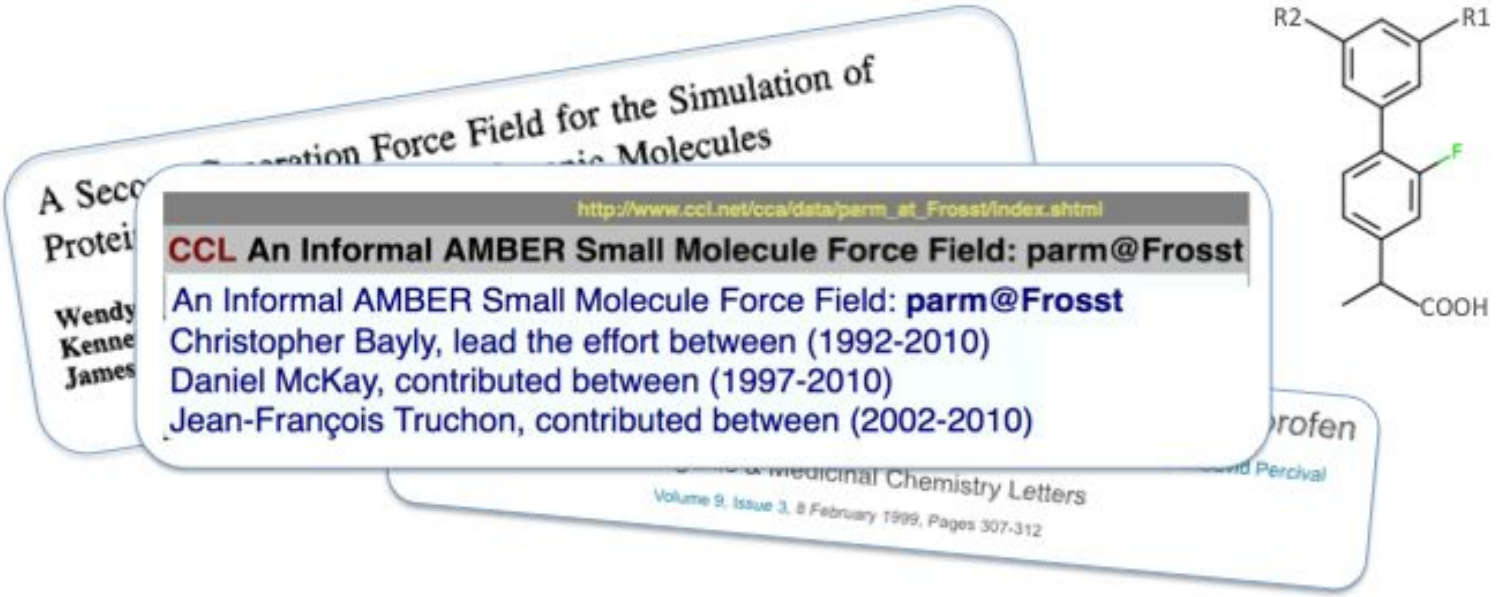
Benjamin D. Sellers,* Natalie C. James, and Alberto Gobbi

Department of Discovery Chemistry, Genentech, Inc., 1 DNA Way, South San Francisco, California 94080, United States

Different force fields might not even agree on the location of a minimum

DOI: 10.1021/acs.jcim.6b00614

parm@frosst is the starting point for a GAFF-like small molecule force field using SMARTS/SMIRKS



Bayly et al.’s parm@Frosst is an AMBER-family small molecule force field and sibling of GAFF

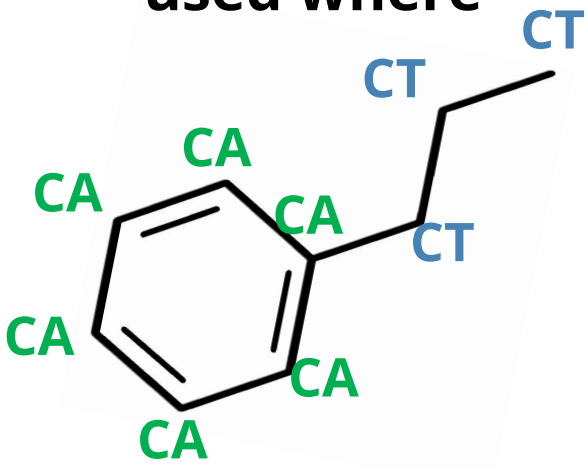
So, how would we use SMIRKS for a force field? Let's think of a carbon-carbon single bond

[#6:1]-[#6:2], length=1.526 angstroms, force constant=620.0 kcal/(mol angstrom^2)

Or, maybe we'd want a generic carbon-carbon bond:
[#6:1]~[#6:2] with its own parameters

Perhaps a more specialized bond?
[#6X3:1]=[#6X3:2] with different parameters

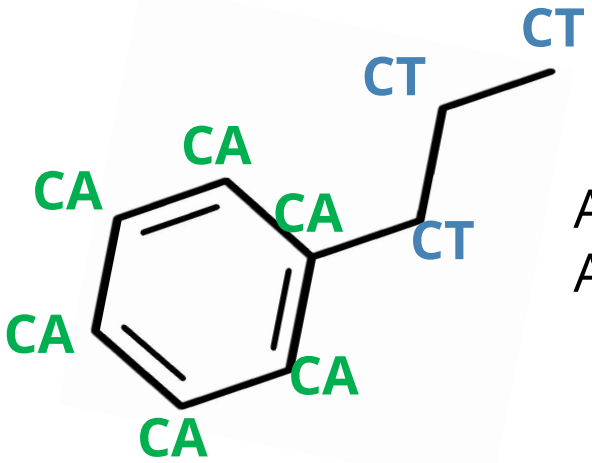
Why is this a good thing? Let's think of atom typing or "chemical perception" which defines which parameters are used where



Aliphatic sp3 carbon (**CT**)
Aromatic sp2 carbon (**CA**)

X -CT-CT-X	Low Barrier Torsion
X-CT-CA-X	Low Barrier Torsion
X -CA-CA-X	High Barrier Torsion

Today's force fields mostly use indirect chemical perception



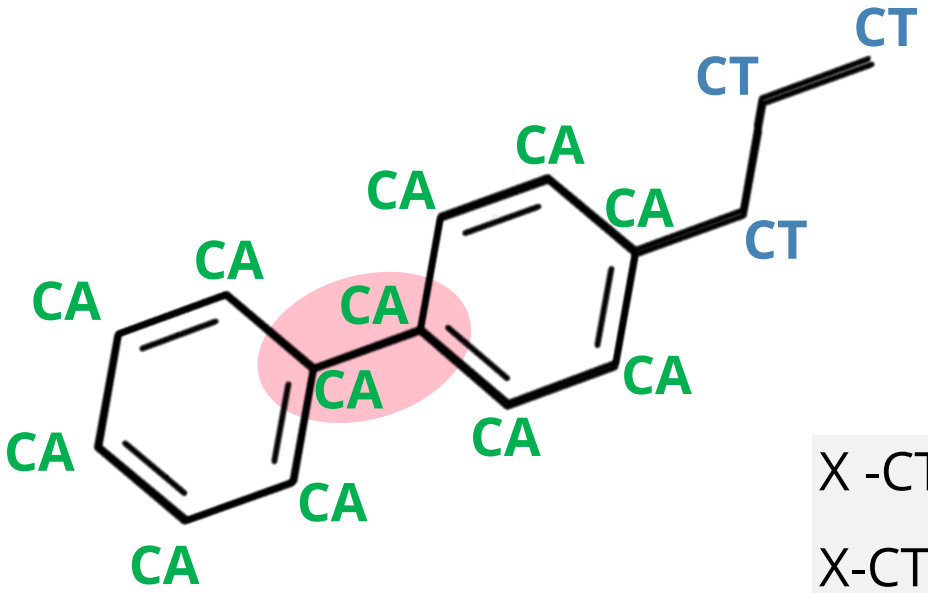
Aliphatic sp3 carbon (**CT**)
Aromatic sp2 carbon (**CA**)

Some tool (or human) assigns atom types

From the atom types, parameters are assigned

Thus, atom types must encode all requisite chemistry and can't be fitted as part of the process

This is a vital issue: Failing to capture the requisite chemistry leads to disaster



Aliphatic sp3 carbon (**CT**)
Aromatic sp2 carbon (**CA**)

X -CT-CT-X	Low Barrier Torsion
X-CT-CA-X	Low Barrier Torsion
X -CA-CA-X	High Barrier Torsion

One can fix this with more complex atom typing

Aliphatic sp3 carbon (**CT**)
Aromatic sp2 carbon (**CA**)
Aromatic carbon bridging phenyl rings (**CP**)

X -CT-CT-X	Low Barrier Torsion
X-CT-CA-X	Low Barrier Torsion
X-CP-CP-X	Low Barrier Torsion
X -CA-CA-X	High Barrier Torsion

Though, new chemistry can still pose challenges so you need an atom type for every special case

Aliphatic sp3 carbon (**CT**)
Aromatic sp2 carbon (**CA**)
Aromatic carbon bridging phenyl rings (**CP**)

Same as CP for multiple bridges (**CQ**)

X -CT-CT-X	Low Barrier Torsion
X-CT-CA-X	Low Barrier Torsion
X-CP-CP-X	Low Barrier Torsion
X -CA-CA-X	High Barrier Torsion
X -CA-CA-X	High Barrier Torsion

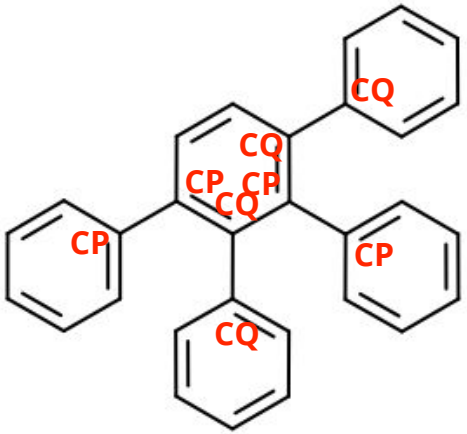
This produces many redundant parameters

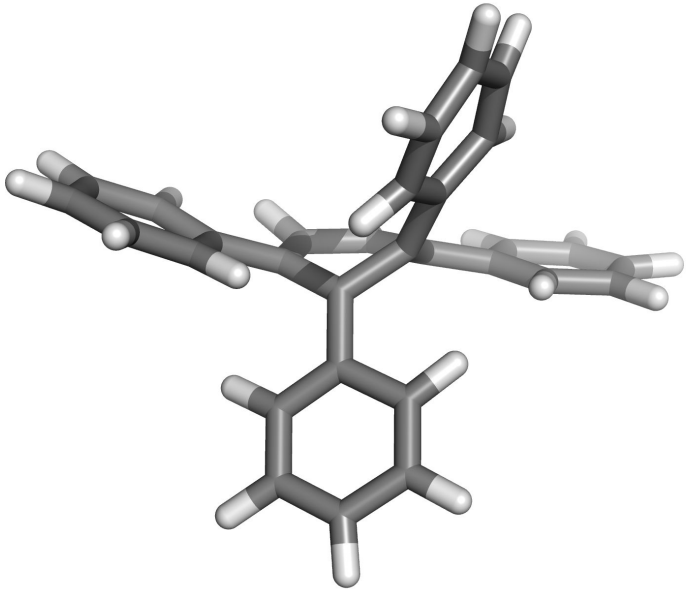
```
CA-CA-SO     70.000 120.000 force ff94 CA-CA-CT
CA-CA-SH     70.000 120.000 std aromatic
CA-CA-SD     70.000 120.000 std aromatic
CA-CA-S      70.000 120.000 std aromatic
CA-CA-P      70.000 120.000 std aromatic
CA-CA-OS     70.000 120.000 ** Gro, JACS,V111,2152('89)
CA-CA-OH     70.000 120.000 ff94 CA-C-OH
CA-CA-O2     70.000 120.000 guess March 5 2009 anionic O
CA-CA-NL     70.000 120.000 guess
CA-CA-ND     70.000 120.000 calc B3PW91/6-31+G** Jan 30 2002
CA-CA-NC     70.000 120.000 quinoline, ff94 CA-CA-CT
CA-CA-NB     70.000 120.000 guess april 11 2000
CA-CA-NA     70.000 120.000 ff94 CA-CA-CT
CA-CA-N3     70.000 120.000 guess april 11 2000
CA-CA-N2     70.000 120.000 ff94 CA-CA-CT
CA-CA-N*     70.000 120.000 ff94 std aromatic
CA-CA-N      70.000 120.000 ff94 CA-CA-CT cb 6jan97
CA-CA-I      70.000 120.000 std sp2 carbon aug 15 2001
CA-CA-F      70.000 120.000 ff94 CA-C-OH
CA-CA-Cl     70.000 120.000 ff94 CA-C-OH
CA-CA-CW     70.000 120.000 amidopyridine, ff94 CA-CA-CT
CA-CA-CR     70.000 120.000 amidopyridine, ff94 CA-CA-CT
```

parm@frosst has a few hundred lines of this type of redundancy

More than 60 identical parameters for CP alone

Downstream problems persist to this day, even in GAFF and GAFF2





- Torsions within the ring end up getting X-CP-CP-X values (rotatable single bond) rather than X-CA-CA-X

Ditching “atom types” for SMIRKS (“parameter types”) allows considerable simplification

For example, GAFF2 has
16 vdW types for carbon

c	1.8606	0.0988
cs	1.8606	0.0988
ca	1.8606	0.0988
cc	1.8606	0.0988
cd	1.8606	0.0988
ce	1.8606	0.0988
cf	1.8606	0.0988
cp	1.8606	0.0988
cq	1.8606	0.0988
cz	1.8606	0.0988
cu	1.8606	0.0988
cv	1.8606	0.0988
cg	1.9525	0.1596
ch	1.9525	0.1596
cx	1.9069	0.1078
cy	1.9069	0.1078

But this should be
three SMIRKS strings

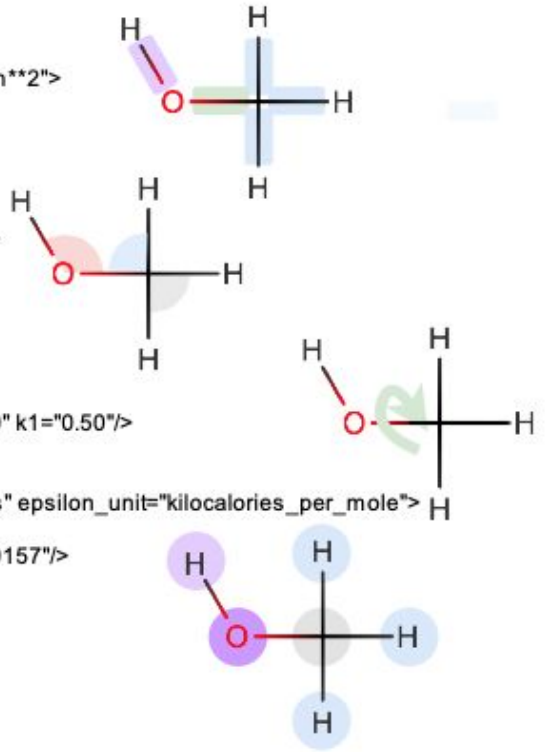
[#6:1]	1.8606	0.0988
[#6X1:1]	1.9525	0.1596
[#6X3r3,#6X3r4:1]	1.9069	0.1078

Very relevant when attempting to automatically fit parameters — are there 32 parameters here, or 6?

(We would argue 6 — the atom types were introduced because of the need for angle or torsional complexity, usually)

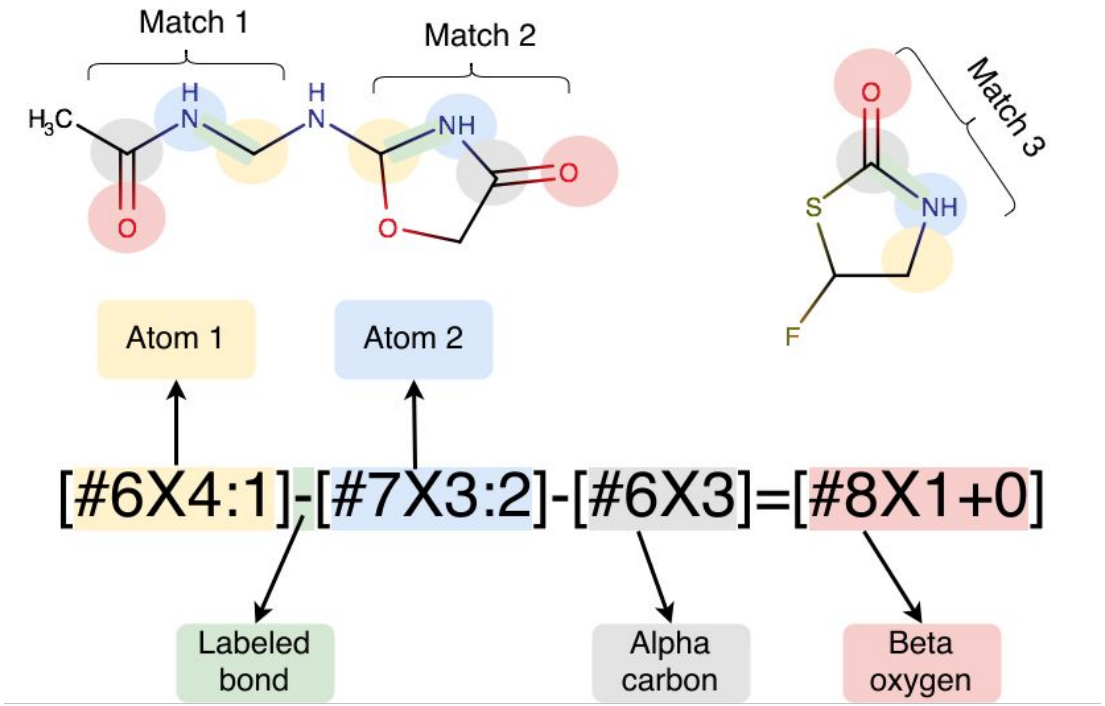
SMIRNOFF parameters for methanol are simple

```
<?xml version="1.0"?>
<SMIRNOFF>

<HarmonicBondForce length_unit="angstroms" k_unit="kilocalories_per_mole/angstrom**2">
  <Bond smirks="[#6X4:1]-[#1:2]" length="1.090" k="680.0"/>
  <Bond smirks="[#6X4:1]-[#8&amp;amp;X2&amp;amp;H1:2]" length="1.410" k="640.0"/>
  <Bond smirks="[#8X2:1]-[#1:2]" length="0.960" k="1106.0"/>
</HarmonicBondForce>

<HarmonicAngleForce angle_unit="degrees" k_unit="kilocalories_per_mole/radian**2">
  <Angle smirks="[a,A:1]-[#6X4:2]-[a,A:3]" angle="109.50" k="100.0"/>
  <Angle smirks="[#1:1]-[#6X4:2]-[#1:3]" angle="109.50" k="70.0"/>
  <Angle smirks="[#6X4:1]-[#8X2:2]-[#1:3]" angle="108.50" k="110.0"/>
</HarmonicAngleForce>

<PeriodicTorsionForce phase_unit="degrees" k_unit="kilocalories_per_mole">
  <Proper smirks="[a,A:1]-[#6X4:2]-[#8X2:3]-[#1:4]" idivf1="3" periodicity1="3" phase1="0.0" k1="0.50"/>
</PeriodicTorsionForce>

<NonbondedForce coulomb14scale="0.833333" lj14scale="0.5" sigma_unit="angstroms" epsilon_unit="kilocalories_per_mole">
  <Atom smirks="[#1:1]" rmin_half="1.4870" epsilon="0.0157"/>
  <Atom smirks="[$([#1]-[#6]-[#7,#8,#9,#16,#17,#35]):1]" rmin_half="1.3870" epsilon="0.0157"/>
  <Atom smirks="[#1$(*-[#8]):1]" rmin_half="0.0000" epsilon="0.0000"/>
  <Atom smirks="[#6:1]" rmin_half="1.9080" epsilon="0.1094"/>
  <Atom smirks="[#8:1]" rmin_half="1.6837" epsilon="0.1700"/>
  <Atom smirks="[#8X2+0$(*-[#1]):1]" rmin_half="1.7210" epsilon="0.2104"/>
</NonbondedForce>

</SMIRNOFF>
```



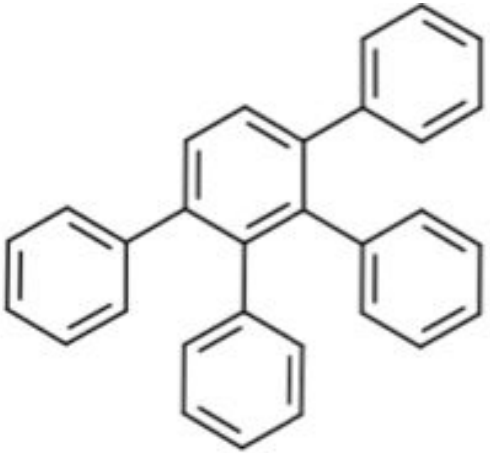
Using SMARTS/SMIRKS allows us to escape atom typing

We use substructure searches on the molecule to assign parameters, rather than atom typing

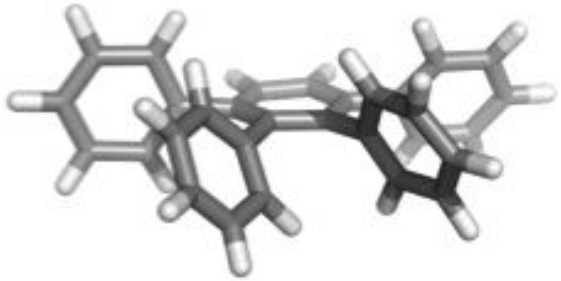


https://github.com/openforcefield/openforcefield/blob/master/The-SMIRNOFF-force-field-format.md

Direct chemical perception utilizes bond order



Torsion Chemistry		Barrier Height	Torsion Minimum
[*:1]~[#6X3:2]:[#6X3:3]~[*:4]	4	14.50	180.0
[*:1]~[#6X3:2]-[#6X3:3]~[*:4]	4	2.50	180.0
…			



We get the geometry right with no special treatment and far fewer parameters

smirnoff99Frosst is our adaptation of parm99+parm@Frosst into this format

Description	Force Field	Lines of parameters
Basic Amber FF:	parm99	720
Merck Frosst small mol:	parm@Frosst	2893
Total:		3613

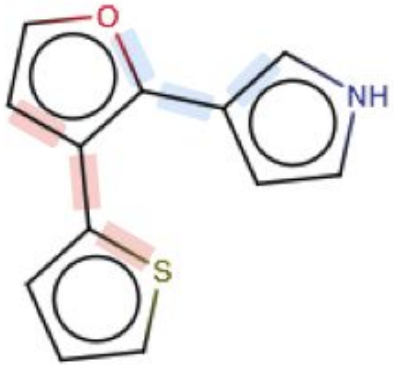
smirnoff99Frosst	332

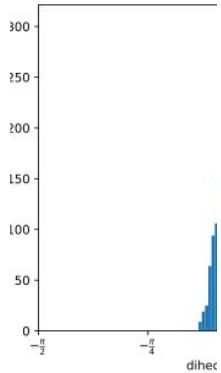
Chemical Space Coverage

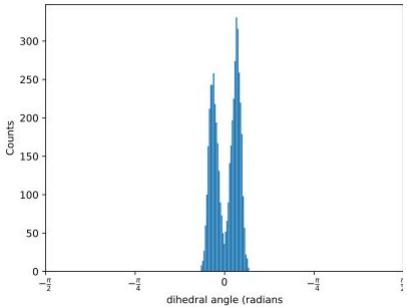
Database	smirnoff 99Frosst	parm @Frosst
DrugBank	99.7%	60%
ZINC	99.8%	52%
eMolecules	99.5%	--

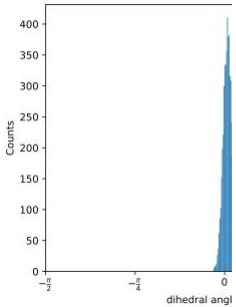
- Less than 1/10 the size of the original force field
- Removes redundancy
- Almost completely covers pharmaceutical chemical space

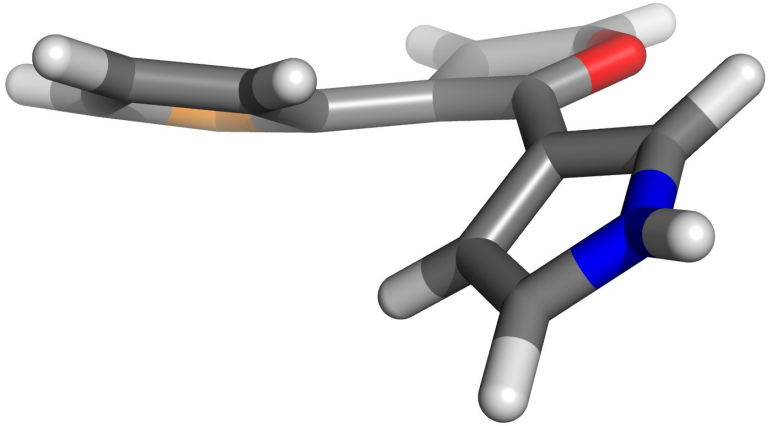
Out of the box it fixes a variety of problems, including siblings of the biphenyl problem





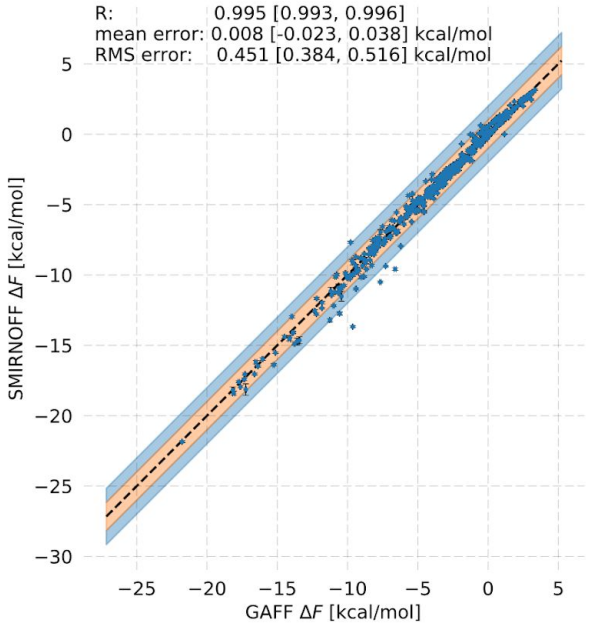


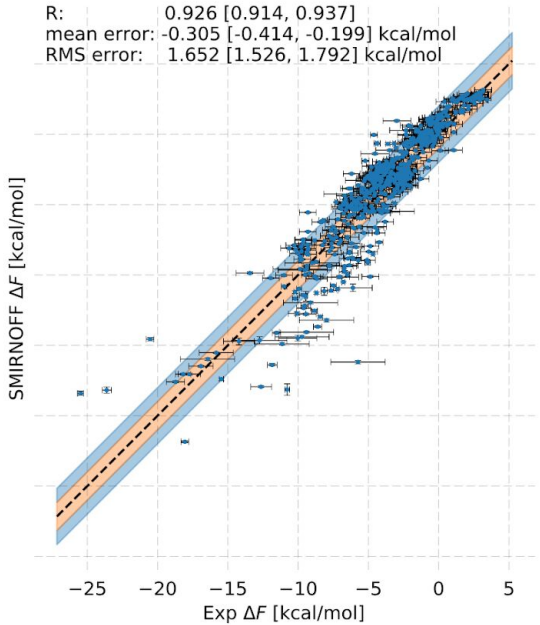


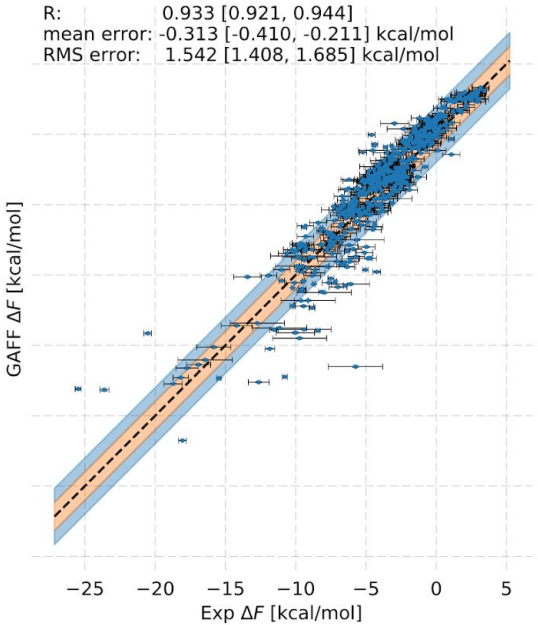


We think smirnoff99Frosst is a great starting point for parameterization work

FreeSolv hydration free energy benchmark





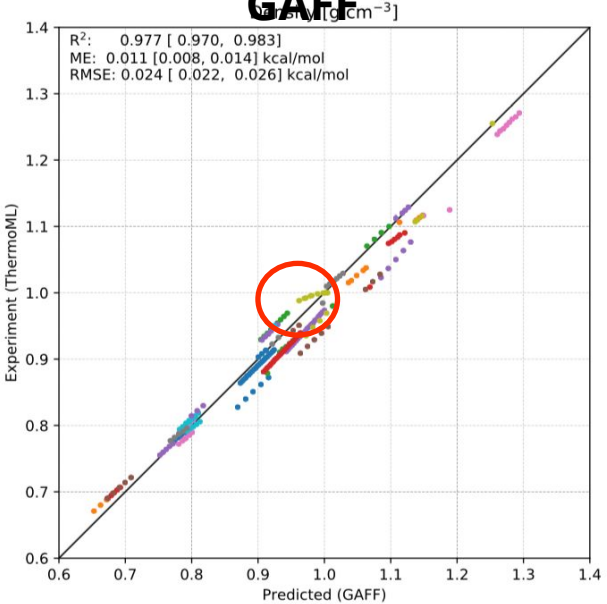


http://github.com/openforcefield/smirnoff99Frosst

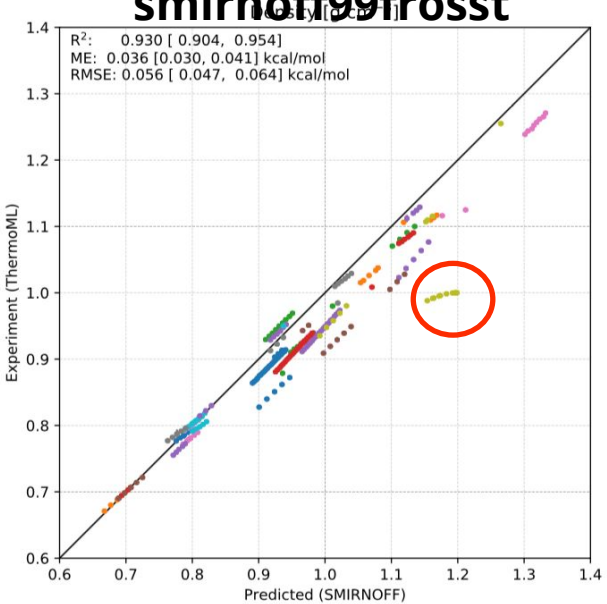
It is competitive with GAFF but with far fewer parameters

ThermoML Archive density benchmark set

GAFF



smirnoff99frosst



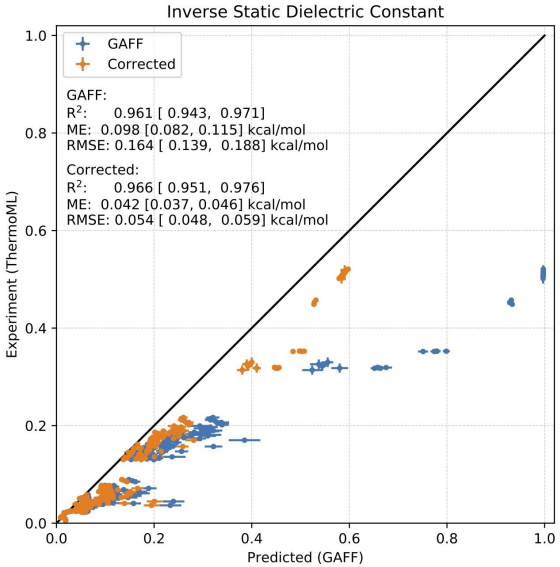
water

http://github.com/openforcefield/smirnoff99Frosst

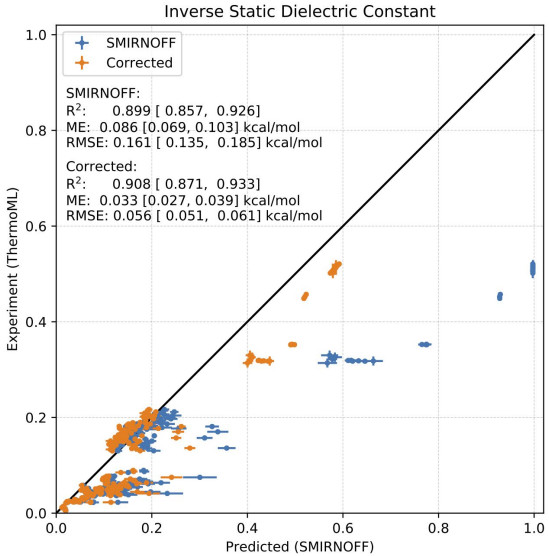
It is competitive with GAFF but with far fewer parameters

ThermoML Archive dielectric benchmark set

GAFF

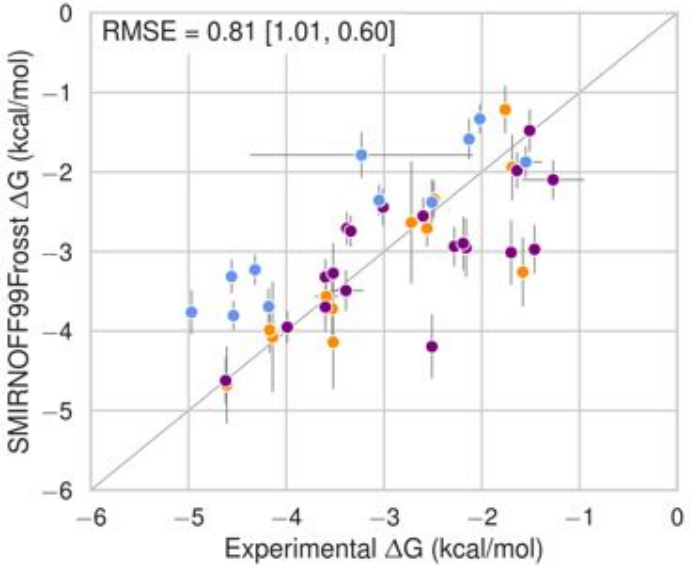


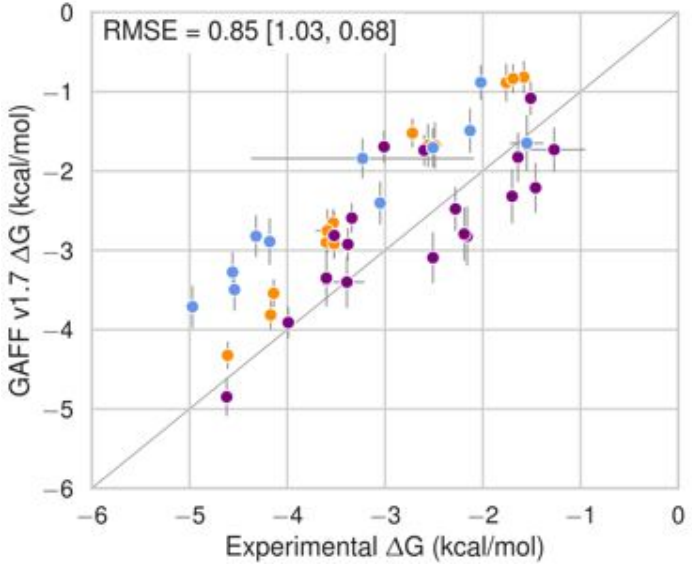
smirnoff99frosst



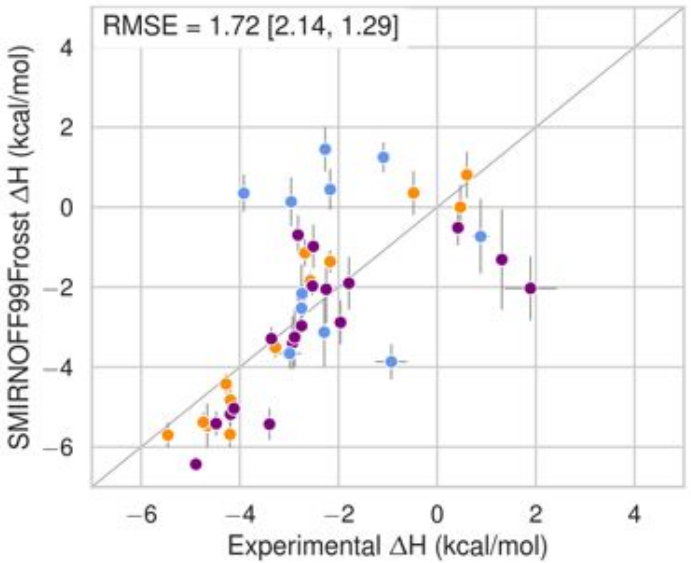
http://github.com/openforcefield/smirnoff99Frosst

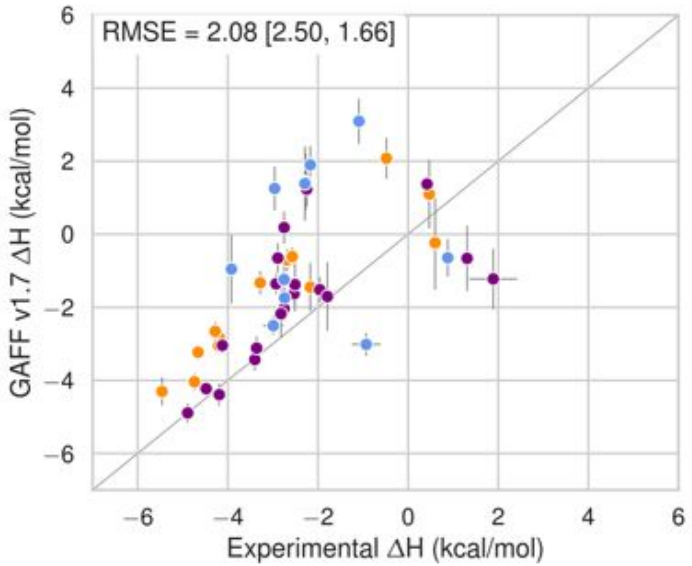
We're also interested in using binding data for fitting, and it turns out to do fairly well on host-guest binding also



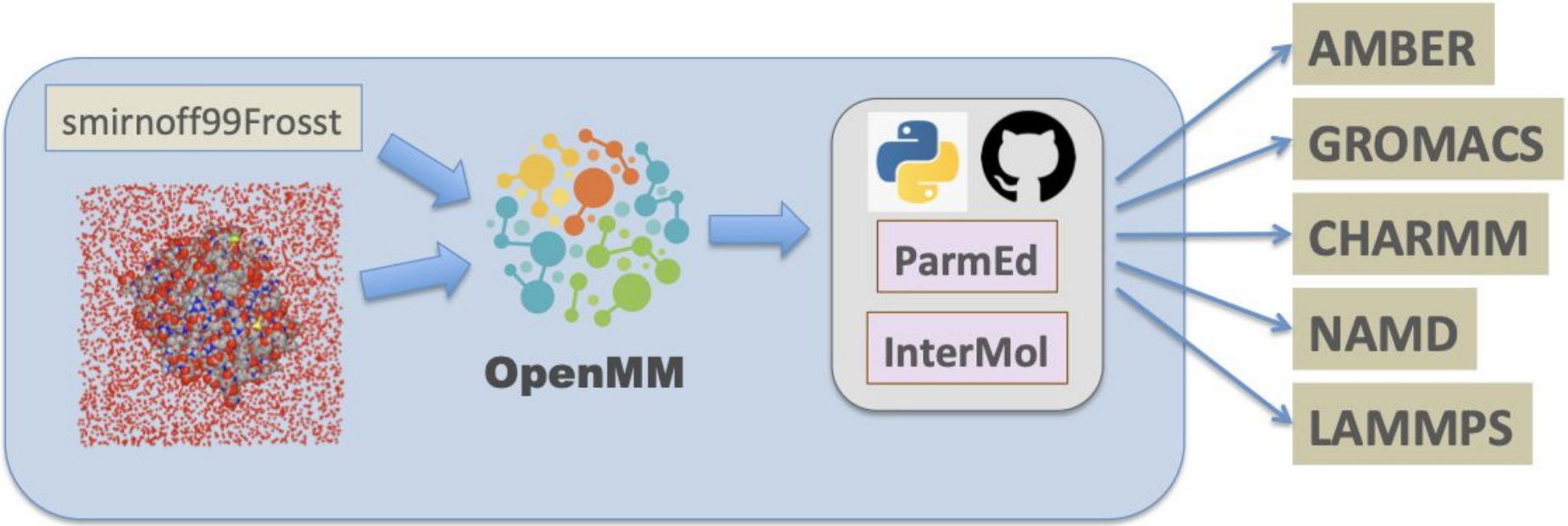


It even works fairly well on enthalpies of binding





You can use smirnoff99Frosst today!

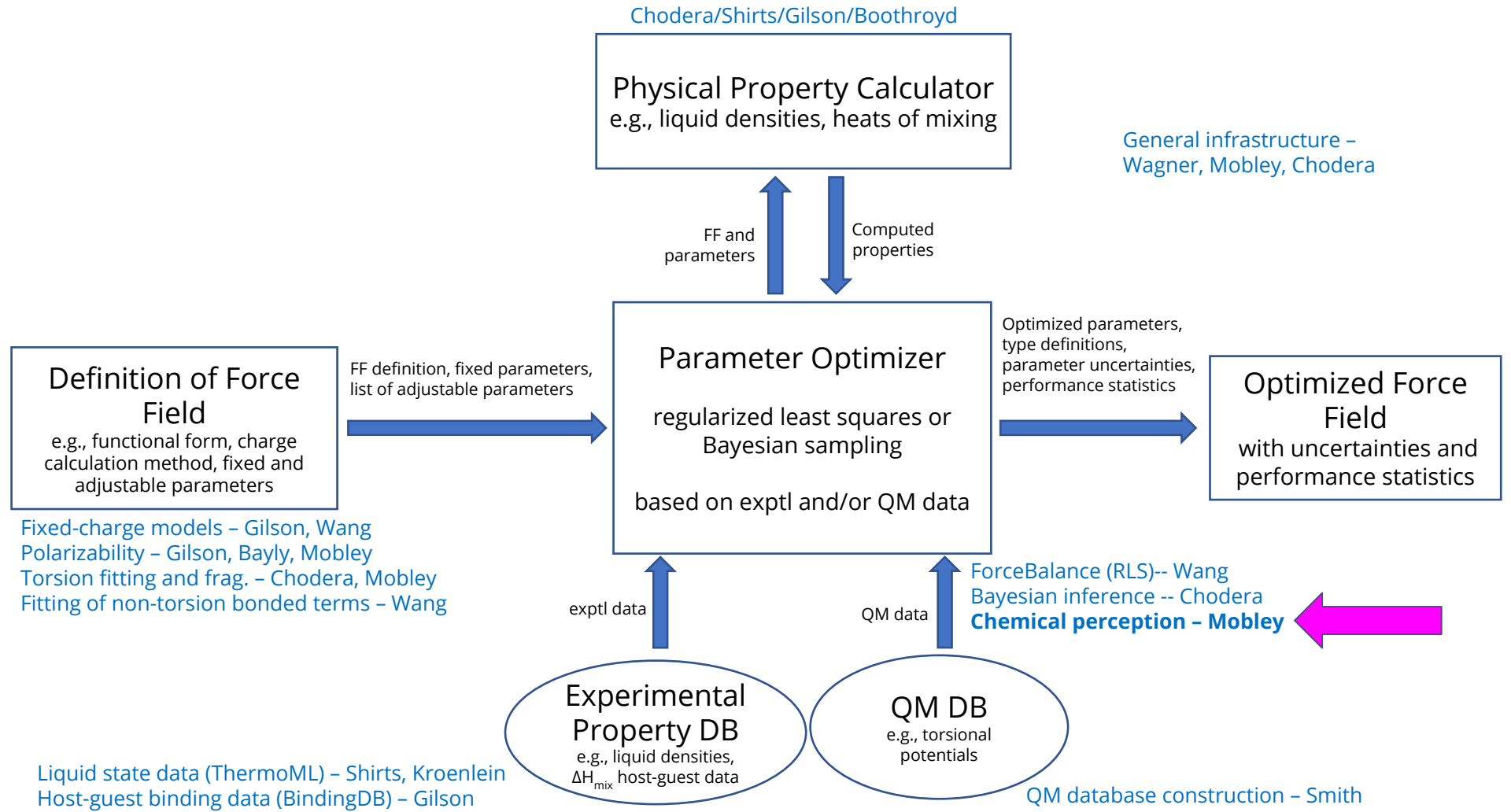


https://github.com/openforcefield/smirnoff99Frosst
https://github.com/openforcefield/openforcefield

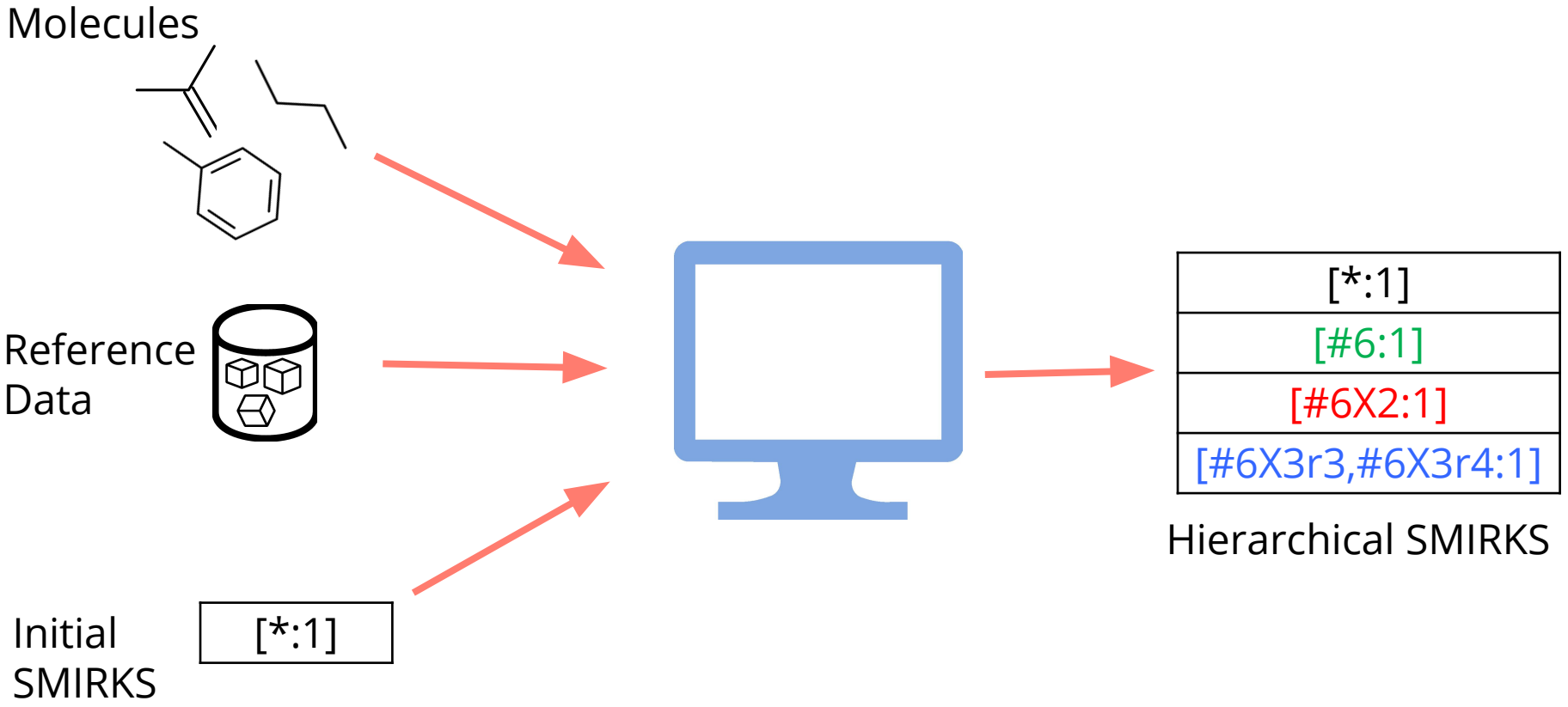
See Slack #onboarding and examples in github.com/openforcefield/openforcefield

Currently, the SMIRNOFF SMIRKS are written by hand

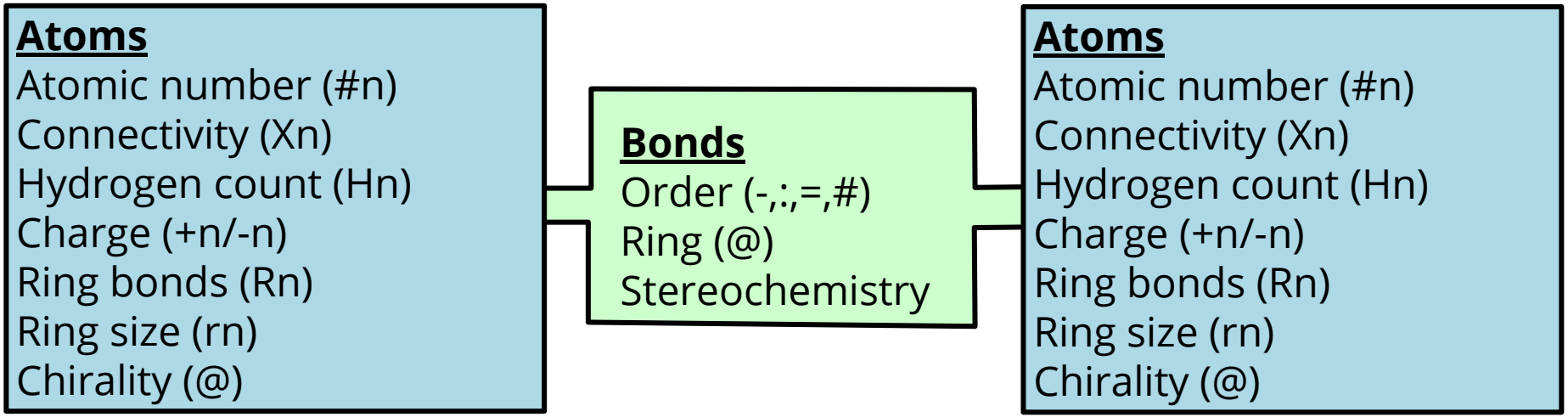
```
"[*:1]~[#6:2]-[#6:3]~[*:4]”      Low Barrier Torsion
"[*:1]~[#6X3:2]:[#6X3:3]~[*:4]”  High Barrier Torsion
```


Chodera/Shirts/Gilson/Boothroyd
Physical Property Calculator
e.g., liquid densities, heats of mixing
General infrastructure –
Wagner, Mobley, Chodera
FF and parameters
Computed properties
Definition of Force Field
e.g., functional form, charge calculation method, fixed and adjustable parameters
FF definition, fixed parameters, list of adjustable parameters
Parameter Optimizer
regularized least squares or Bayesian sampling
based on exptl and/or QM data
Optimized parameters, type definitions, parameter uncertainties, performance statistics
Optimized Force Field
with uncertainties and performance statistics
Fixed-charge models – Gilson, Wang
Polarizability – Gilson, Bayly, Mobley
Torsion fitting and frag. – Chodera, Mobley
Fitting of non-torsion bonded terms – Wang
exptl data
QM data
ForceBalance (RLS)-- Wang
Bayesian inference -- Chodera
Chemical perception – Mobley
Experimental Property DB
e.g., liquid densities, ΔH_mix host-guest data
QM DB
e.g., torsional potentials
Liquid state data (ThermoML) – Shirts, Kroenlein
Host-guest binding data (BindingDB) – Gilson
QM database construction – Smith

Chemical perception choices should be data-driven



SMIRKS have simple components, but are a complex language

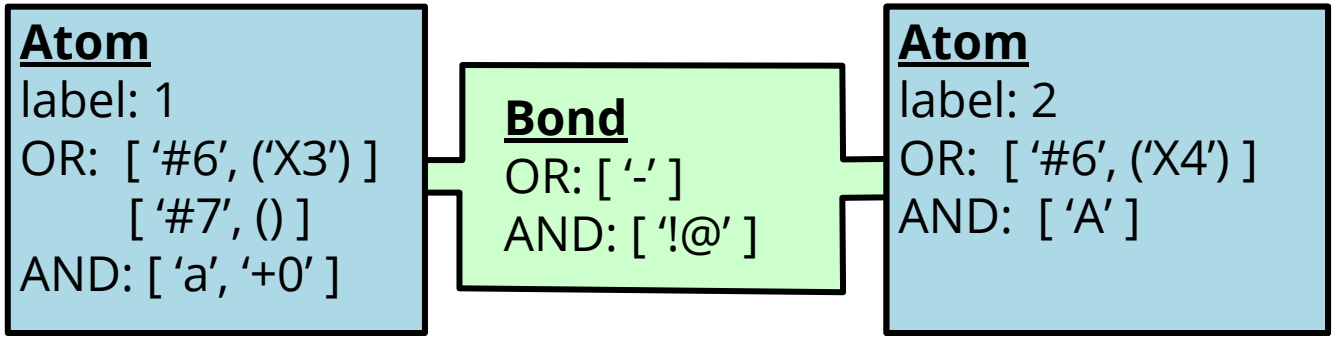


“[#6,#“[C,N]-;@[C,N]”;;A:2]”

We needed a tool to edit SMIRKS by component

“[#6X3,#7;a;+0:1]-;!@[#6;A:2]” → “[#6X3,#7;a;+0:1]-;!@[#6X4;A:2]”

ChemicalEnvironment objects

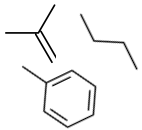


- store SMIRKS information by atom and bond
- allow for decoration changes
- are input and output as SMIRKS patterns

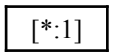
SMIRKY: a Monte Carlo algorithm for learning SMIRKS in a SMIRNOFF

1. Chose a random SMIRKS
2. (a) Create a new SMIRKS by randomly changing this or (b) Delete this SMIRKS
3. Compute a change in “score” based on similarity to an existing FF
4. Use the Metropolis criterion to accept or reject the proposed move

Reference Data: SMIRNOFF99Frosst



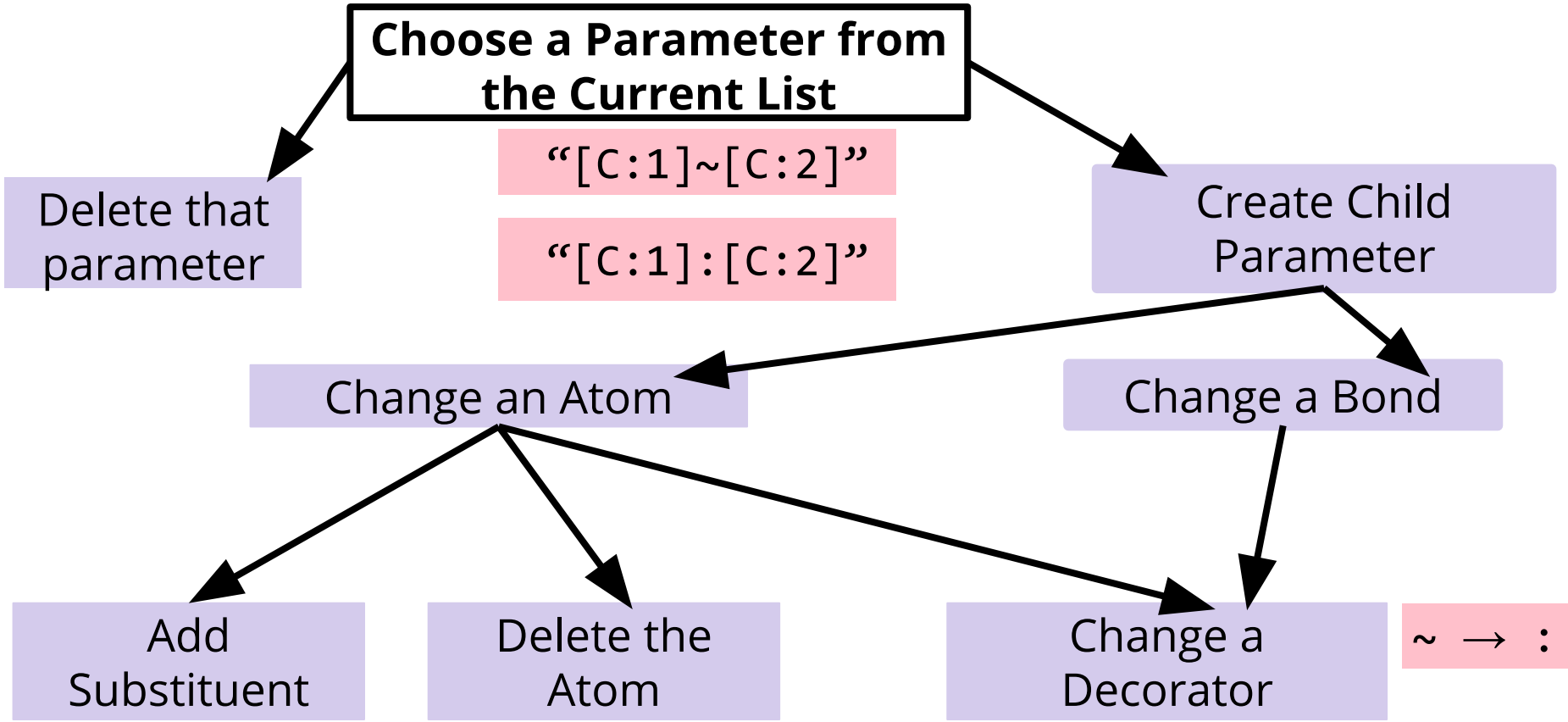
Molecules: 3 sets of increasing chemical complexity



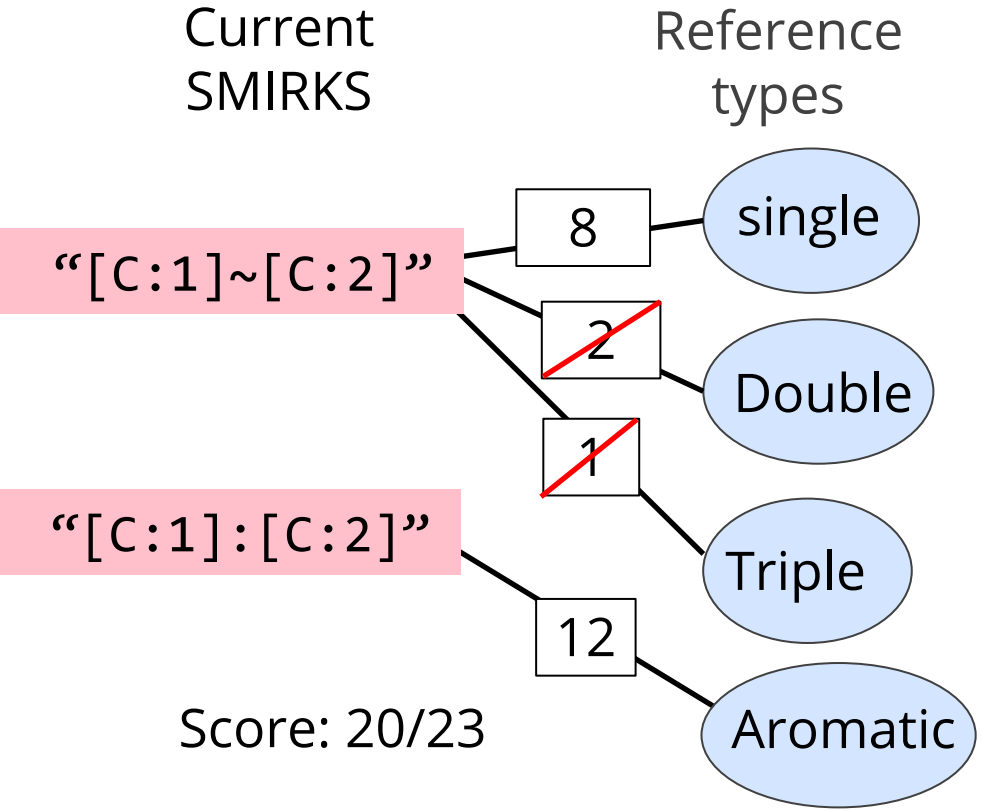
Initial SMIRKS: Simple based on parameter type

Zanette, et al. doi:10.1021/acs.jctc.8b00821

We created a move set to create SMIRKS for chemical environment



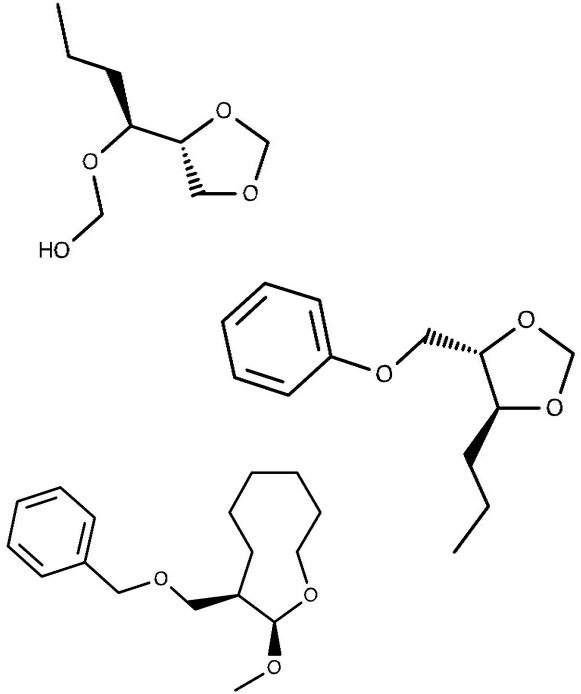
SMIRKS sets are scored based on typed molecules

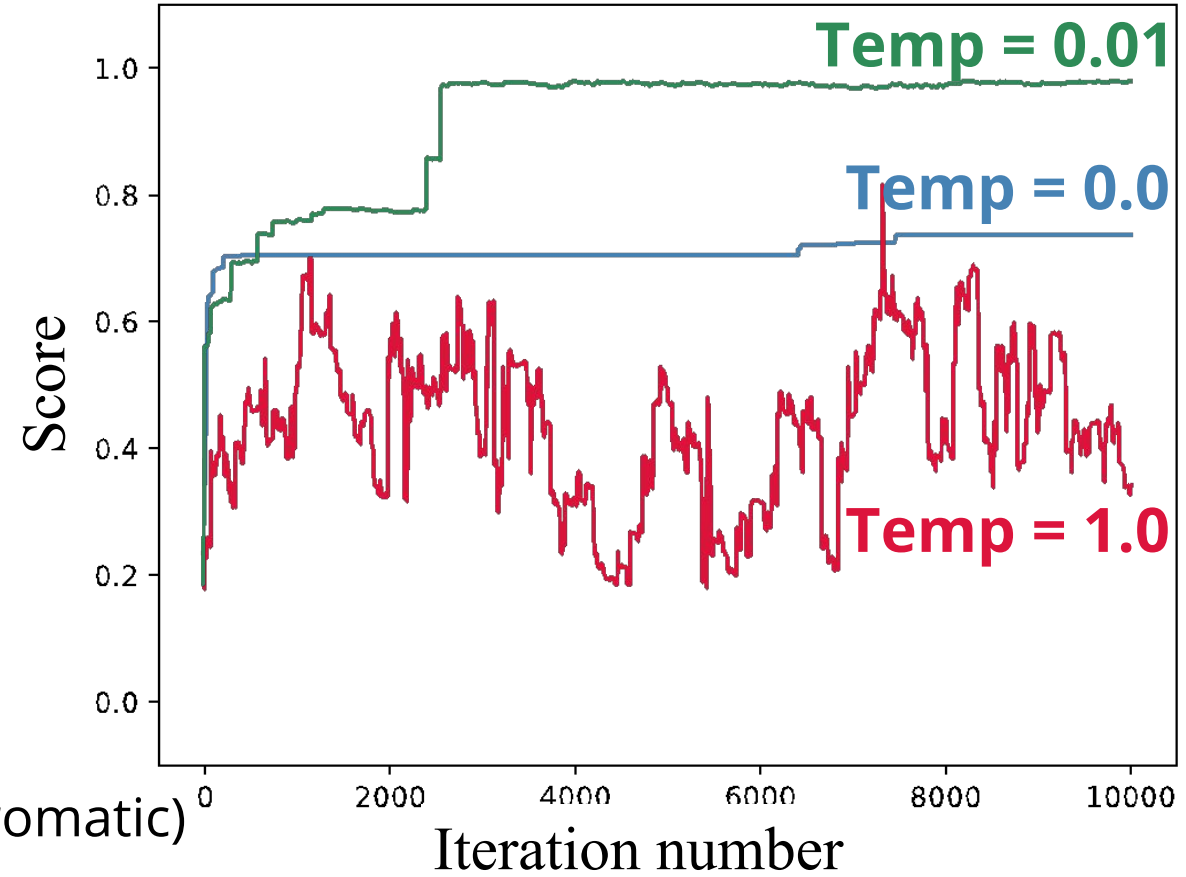


Accepted if $R < \exp\left[\frac{S_{curr} - S_{prev}}{T}\right]$

R = Random[0,1]
S = Score
T = Temperature

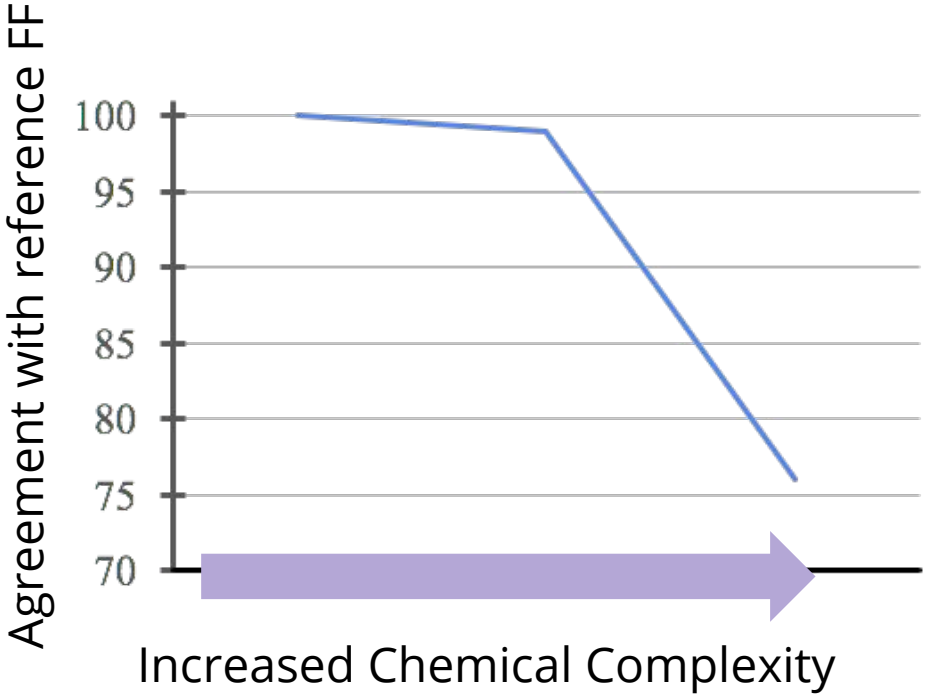
We can find SMIRNOFF parameters in a simple molecule sets





200 molecules with
oxygen, hydrogen,
and carbon (aliphatic and aromatic)

Realistic molecule sets were less successful due to the large combinatorial space



How long to make this Target SMIRKS pattern

[*:1]-,:[#6X3:2]=[#7X2:3]-[*:4]

1. Pick the right starting SMIRKS

[*:1]~[#6:2]~[#7:3]~[*:4]

Probability ~ 0.02

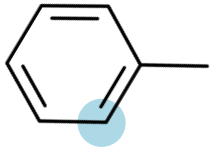
2. Right decorator to the right atom

Probability ~ 0.01

Approximately 1,000,000,000 iterations

Important SMIRKY lessons for moving forward

- Naive moves in SMIRKS space waste time and are unnecessary



#6,X3,H1,+0,R2,r6

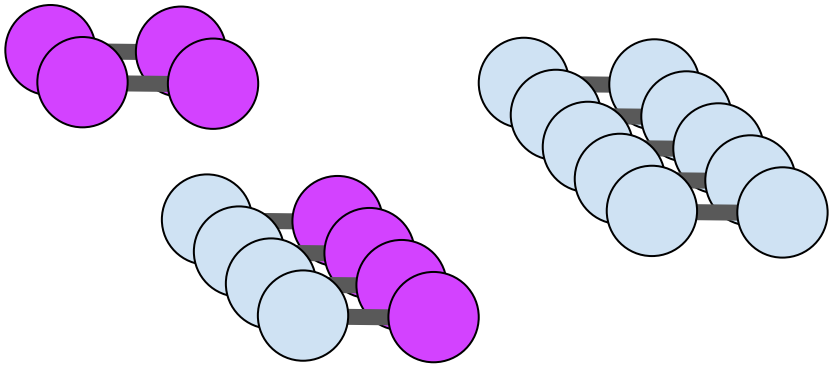
- SMIRKY was built exclusively for comparison with reference force fields

Reference Data and scoring

SMIRKS Generation

Input Molecules

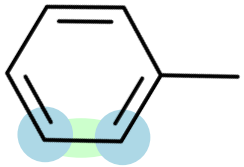
ChemPer extracts decorators to make SMIRKS from input molecular fragment



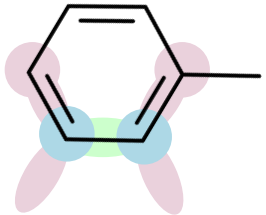
SMIRKS 1
SMIRKS 2
SMIRKS 3

- How does ChemPer create SMIRKS patterns
- Challenges addressed in this package
- How to generate SMIRKS from reference QM data

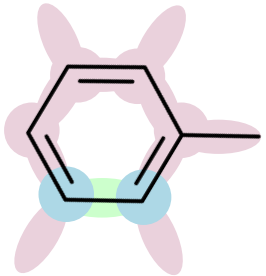
SMIRKS are generated starting with the indexed atoms



'[#6aH1X3x2r6+0:1]:@[#6aH1X3x2r6+0:2]'



'[#6aH1X3x2r6+0:1](-!@[#1AH0X1x0!r+0])(:@[#6aH1X3x2r6+0]):@[#6aH1X3x2r6+0:2](-!@[#1AH0X1x0!r+0]):@[#6aH1X3x2r6+0]'



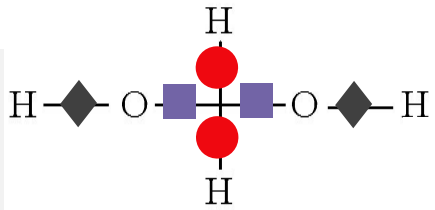
'[#6aH1X3x2r6+0:1](-!@[#1AH0X1x0!r+0])(:@[#6aH1X3x2r6+0](-!@[#1AH0X1x0!r+0]):@[#6aH0X3x2r6+0](-!@[#6AH3X4x0!r+0](-!@[#1AH0X1x0!r+0])(-!@[#1AH0X1x0!r+0])-!@[#1AH0X1x0!r+0]):@[#6aH1X3x2r6+0](-!@[#1AH0X1x0!r+0]):@[#6aH1X3x2r6+0]-!@[#1AH0X1x0!r+0]):@[#6aH1X3x2r6+0:2]-!@[#1AH0X1x0!r+0]'

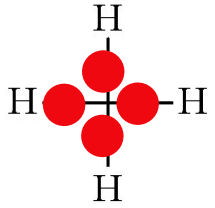
ChemPer: Builds SMIRKS patterns from clusters of molecular fragments

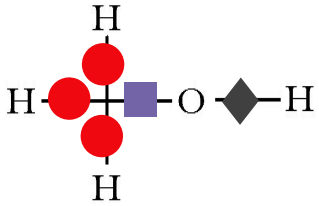
<Bond smirks="[#6X4:1]-[#6X4:2]" k="620.0" length="1.526"/>

<Bond smirks="[#6:1]-[#8:2]" k="640.0" length="1.410"/>

<Bond smirks="[#8:1]-[#1:2]" k="1106.0" length="0.960"/>







[#6AH2X4x0!r+0,#6AH3X4x0!r+0,#6AH4X4x0!r+0:1]-;!@[#1AH0X1x0!r+0:2]

[#6AH2X4x0!r+0,#6AH3X4x0!r+0:1]-;!@[#8AH1X2x0!r+0:2]

[#8AH1X2x0!r+0:1]-;!@[#1AH0X1x0!r+0:2]

These SMIRKS are generalized by removing unnecessary decorators

[#6AH2X4x0!r+0,#6AH3X4x0!r+0,#6AH4X4x0!r+0:1]-;!@[#1AH0X1x0!r+0:2]

[#6AH2X4x0!r+0,#6AH3X4x0!r+0:1]-;!@[#8AH1X2x0!r+0:2]

[#8AH1X2x0!r+0:1]-;!@[#1AH0X1x0!r+0:2]

Iteratively remove unnecessary decorators while maintaining clustering

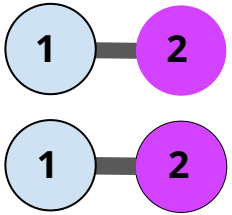
[#6:1]-[#6:2]

[#6:1]-[#8:2]

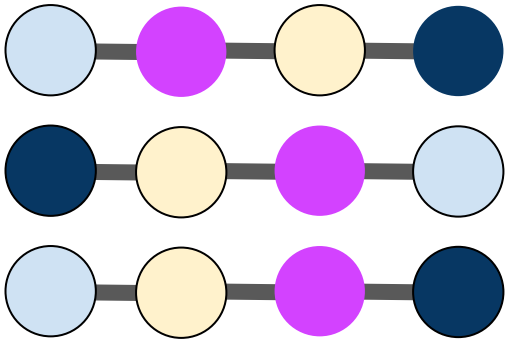
[#8:1]-[#1:2]

Fragments need to be properly aligned

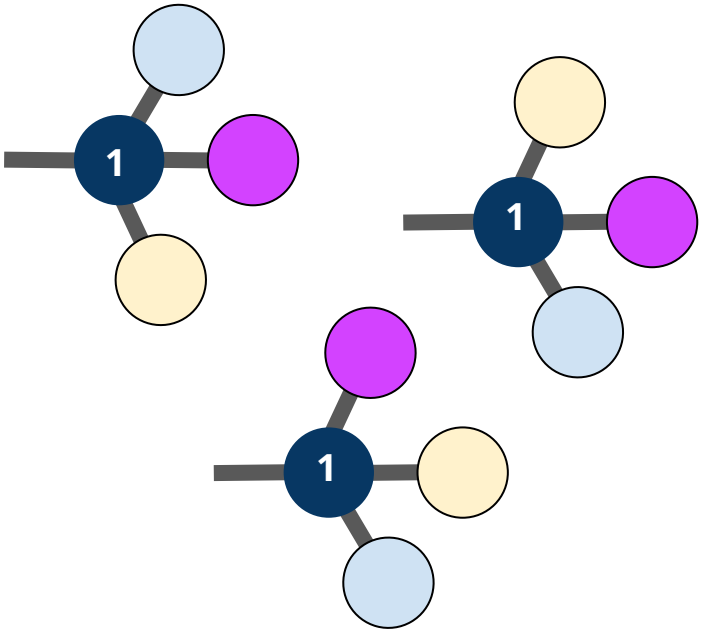
SMIRKS need to be created from aligned indexed atoms



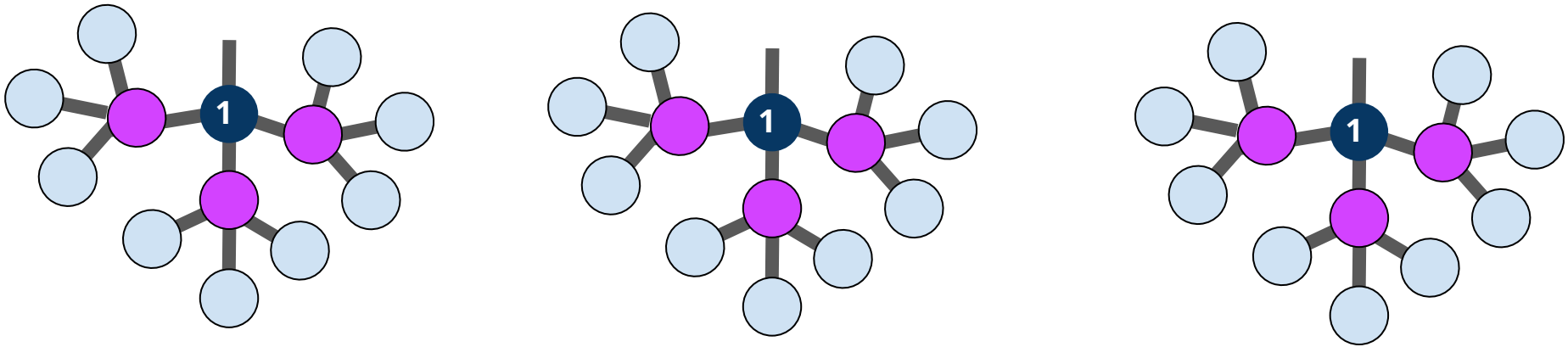
This is slightly more complex for torsions



Just as important, unindexed atoms need to agree when they are added



Size of SMIRKS is determined automatically



- Layers of atoms are added until SMIRKS match the input clustering
- Non-indexed atoms can be removed during SMIRKS reduction

Order is important when creating hierarchical SMIRKS

Handwritten SMIRKS increase in complexity

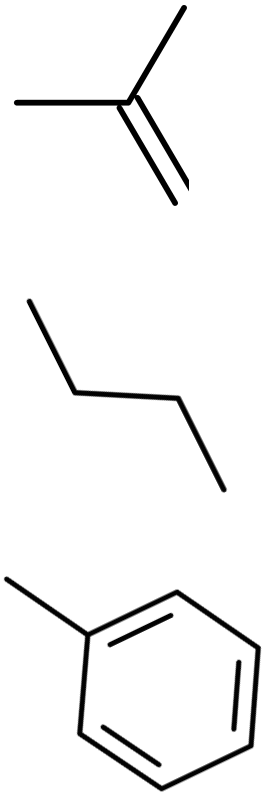
[*:1]
[#6:1]
[#6X2:1]
[#6X3r3,#6X3r4:1]

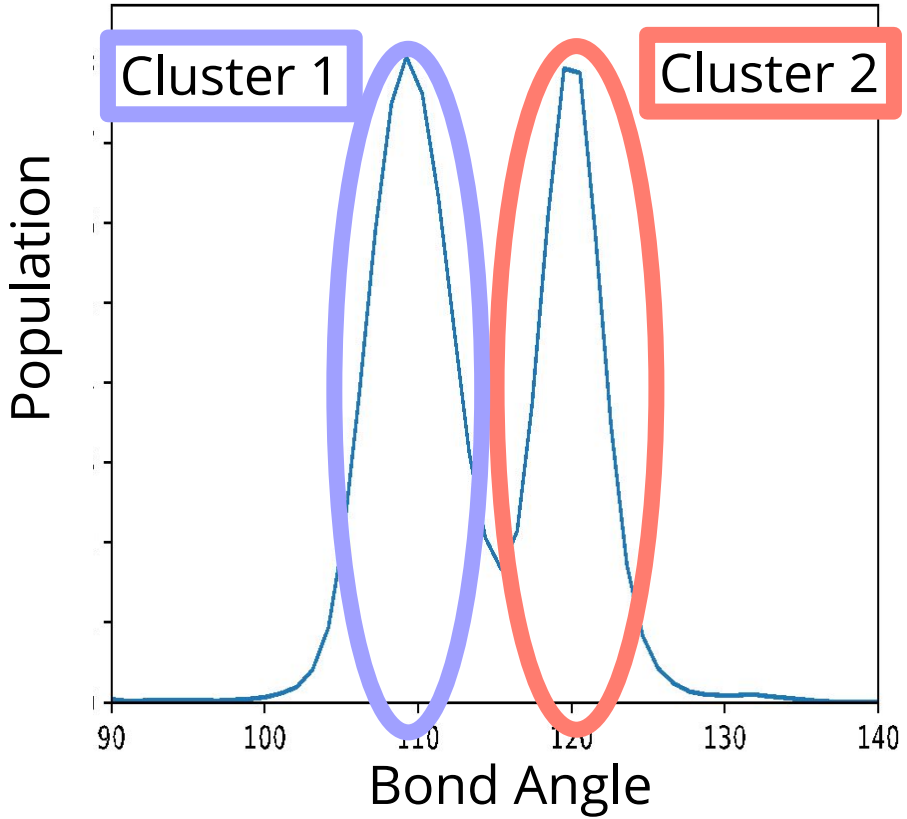
When automatically generating SMIRKS, the “correct” order should be determined by input molecules and reference data.

Sort clusters of fragments by

- Number of fragments in the cluster
- Those with more decorators per atom
- Random, checking multiple options

Toy Example: Consider finding parameters for angles around carbon

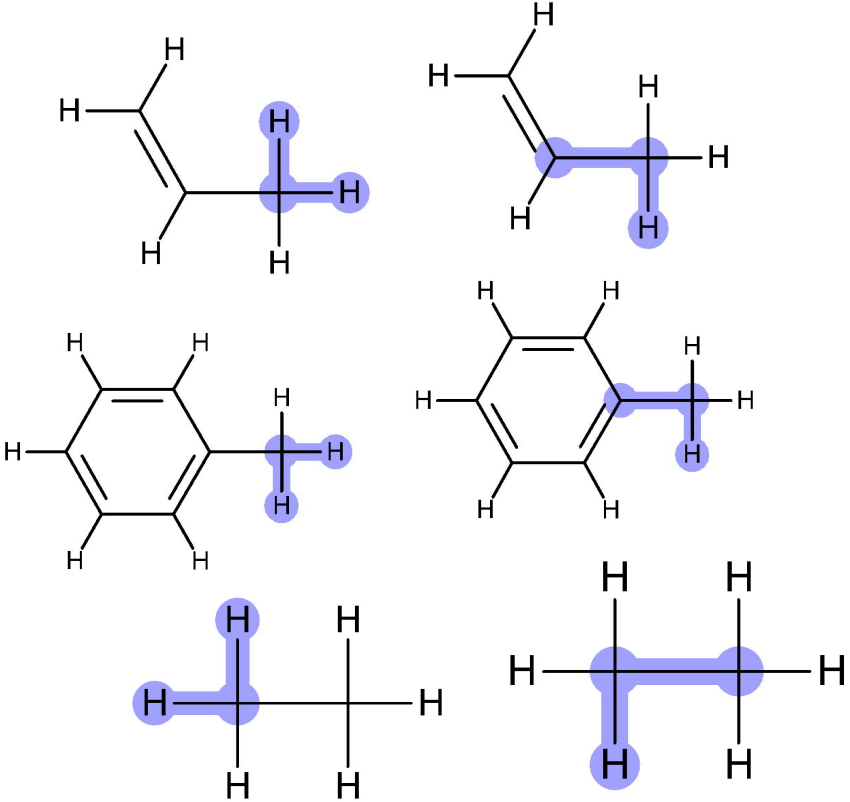




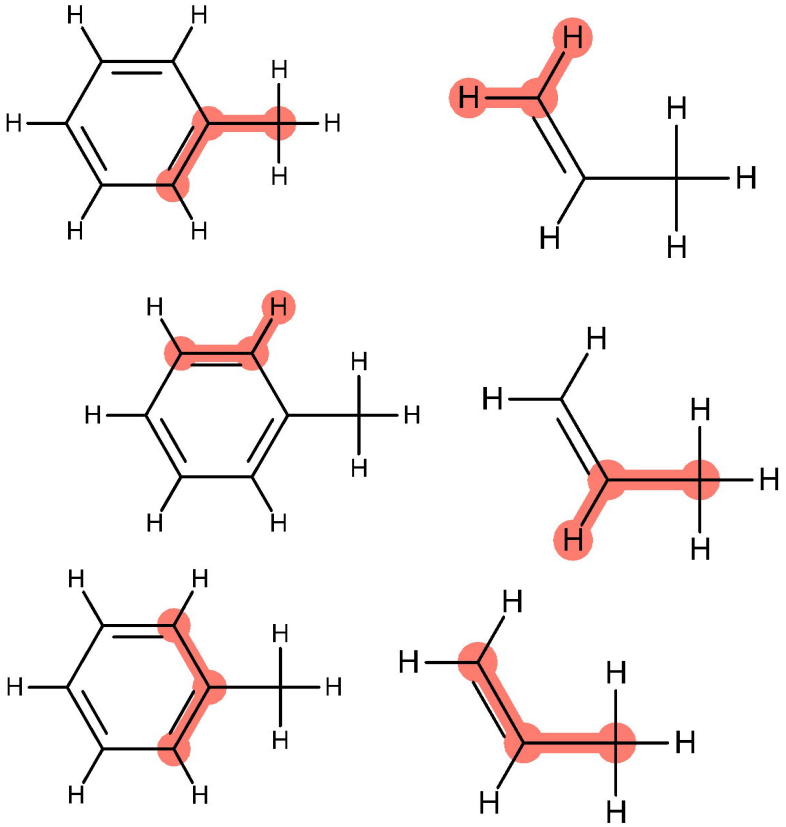
Initially identified two clusters.

Step 1: cluster atoms by which parameter they will be assigned

Cluster 1

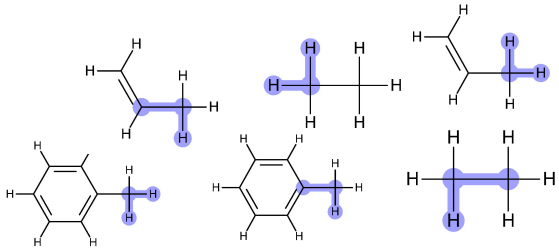


Cluster 2



Step 2: Extract all possible decorator combinations for each cluster

Cluster 1

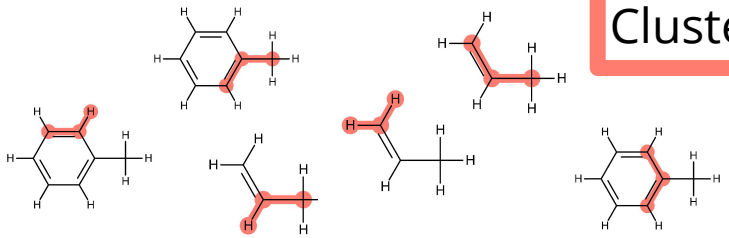


[#1AH0X1x0!r+0,#6AH3X4x0!r+0,#6aH0X3x

[*:1]-[#6X4]-[*:3]

,#6AH0X3x0!r+0,#6AH2X4x0!r+0:3]

Cluster 2



[#1AH0X1x0!r+0,#6AH3X4x0!r+0,#6aH1X3x2r6+0:1]-,: [*:1]-[#6X3]-[*:3]

x2r6+

6AH0X3x0!r+0,#6AH2X3x0!r+0,#6AH3X4x0!r+0,#6aH1X3x2r6+0:3]

Step 3: Systematically reduce SMIRKS to only the essentials

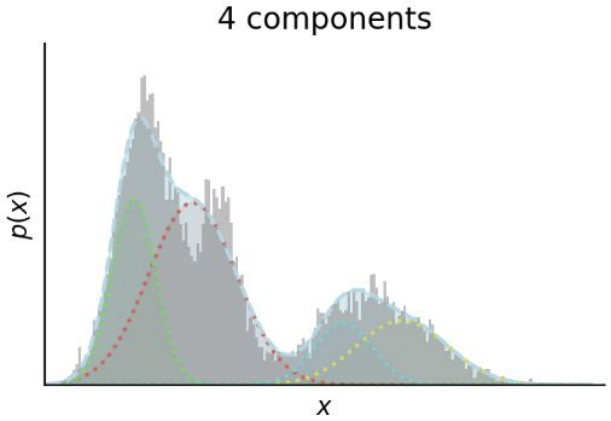
- Compare decorators in each cluster
- Remove all unnecessary decorators while maintaining clustering

QM data for valence terms should drive chemical perception choices

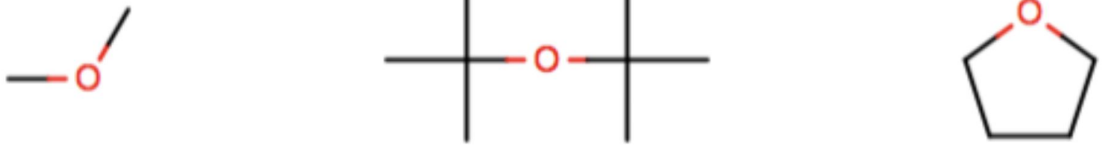
- Bonds and angles
  - Cluster based on bond lengths/angles and/or force constants
  - #Valence
- Proper torsions
  - Fragment molecules and then cluster torsions using force constants
  - #Torsions
- Improper torsion
  - Cluster using phase angles and force constants
  - Starting with nitrogen centers
  - #Improper-torsion

The best types of data and clustering algorithms are still being considered

- ChemPer is modular so we can easily try new clustering algorithms

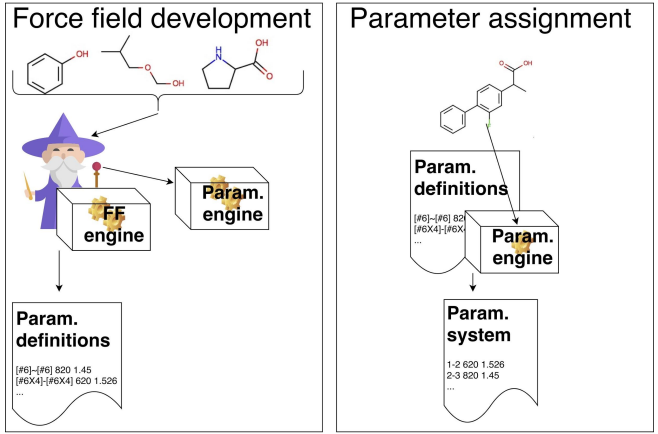


- Concerns about multiple contributions to energy changes in valence terms

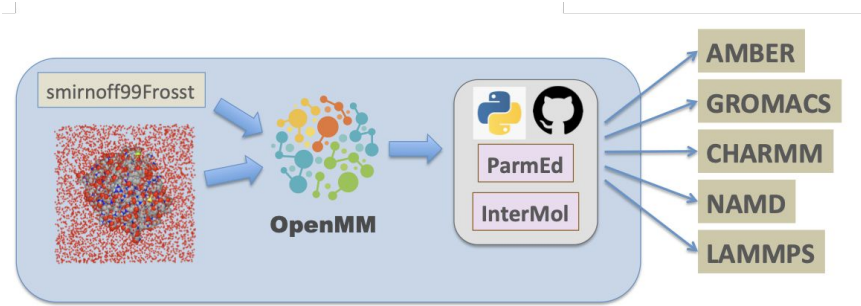


Θ_0:	111.6	125.9	108.5
k:	small	medium	large

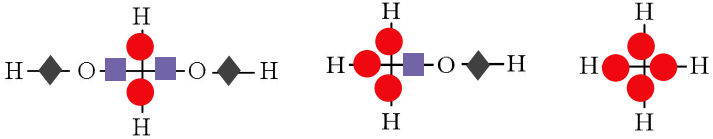
Direct chemical perception reduces redundancy



SMIRNOFF99Frosst is available now



ChemPer makes SMIRKS from clustered fragments



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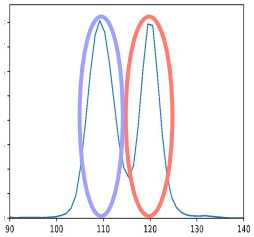
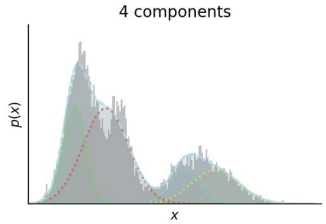
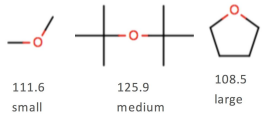
[#6AH2X4x0!r+0,#6AH3X4x0!r+0:1]-;!@[#8AH1X2x0!r+0:2]

[#8AH1X2x0!r+0:1]-;!@[#1AH0X1x0!r+0:2]

Further research into types of clustering data and algorithms is required







Acknowledgements

GitHub:

- github.com/openforcefield/openforcefield
- github.com/openforcefield/smirnoff99Frosst
- github.com/mobleylab/chemper

Slack:

- #smirnoff
- #smarty