

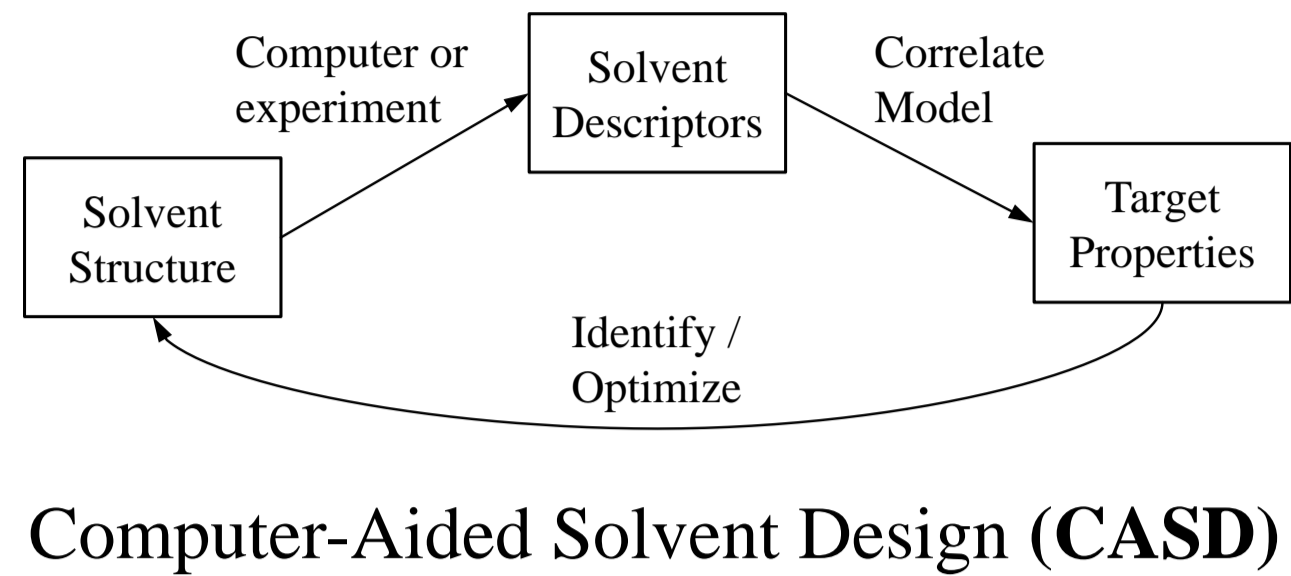
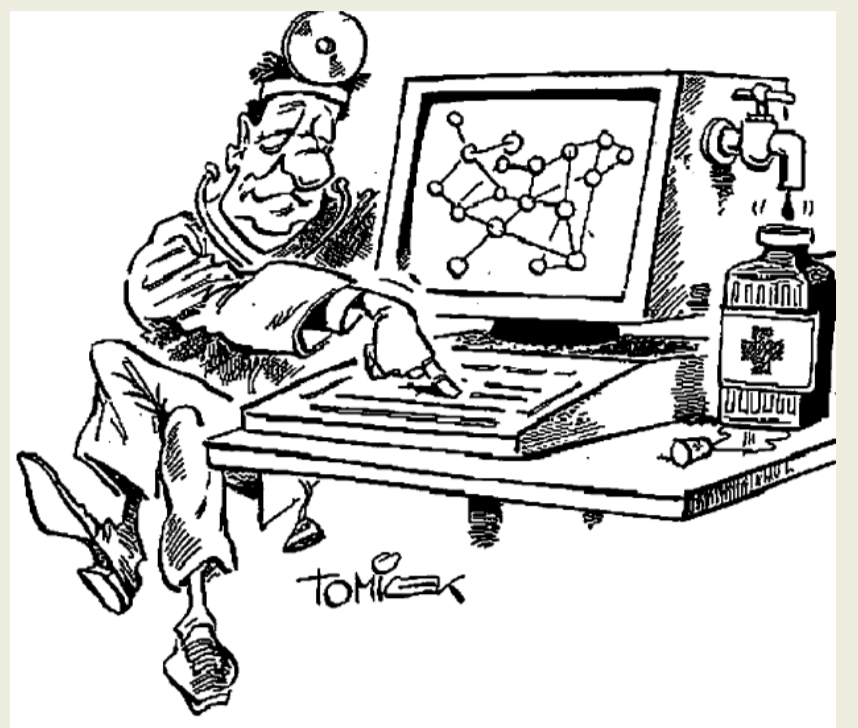


MAX-PLANCK-GESELLSCHAFT

Integrated solvent and process design exemplified for a Diels-Alder reaction

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Research Motivation



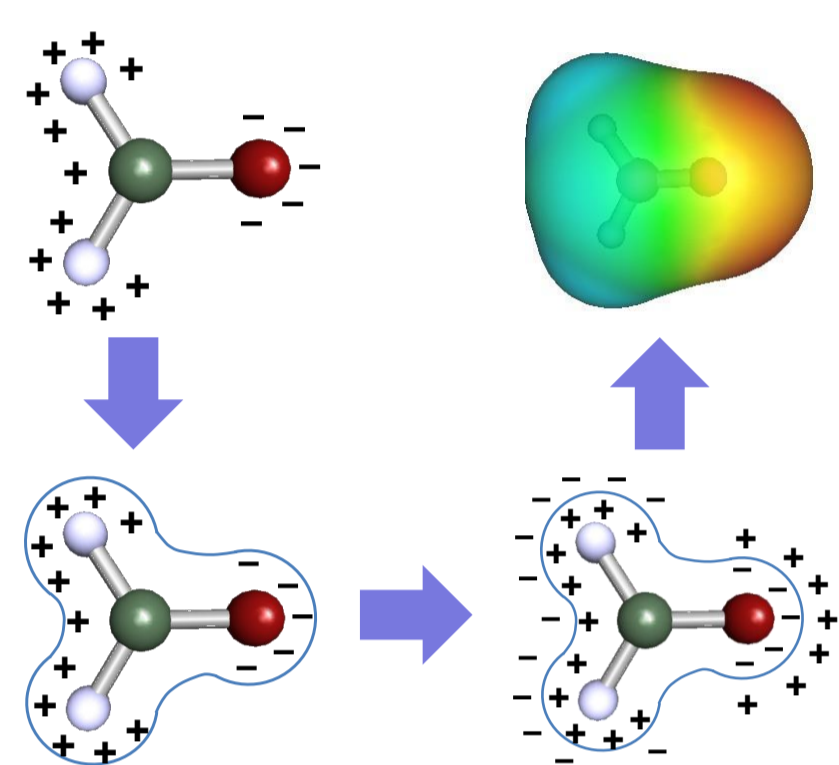
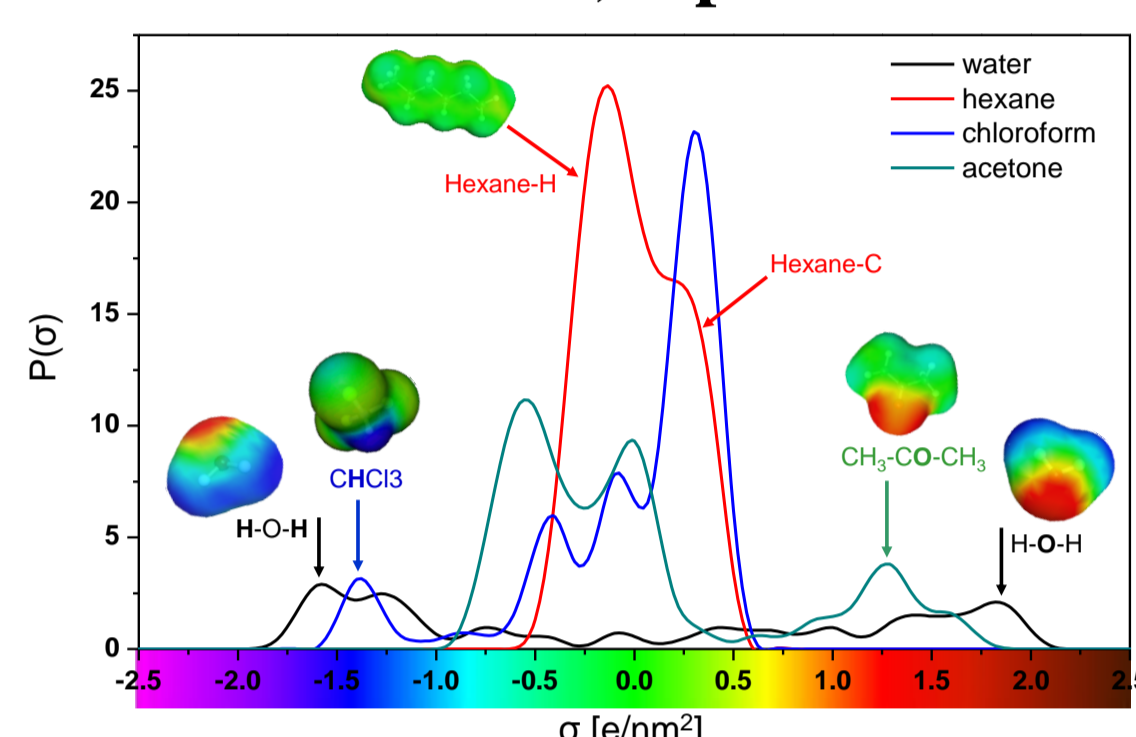
- **CASD:** First relates solvent structure to properties through solvent descriptors, then reversely determines the optimal solvent structure that match target property values based on the established quantitative structure-property relationship (QSPR).
- Conventional QSPR uses empirical or experimentally determined solvent parameters to correlate solvent effects. It can not be directly applied for extensive solvent screening and solvent molecular design.
- Integrated solvent and process design is essential for achieving an optimal chemical process.

Goal: to develop methods for the design of solvents for chemical reactions and integrated reactive processes.

Theoretical Concepts & Methods

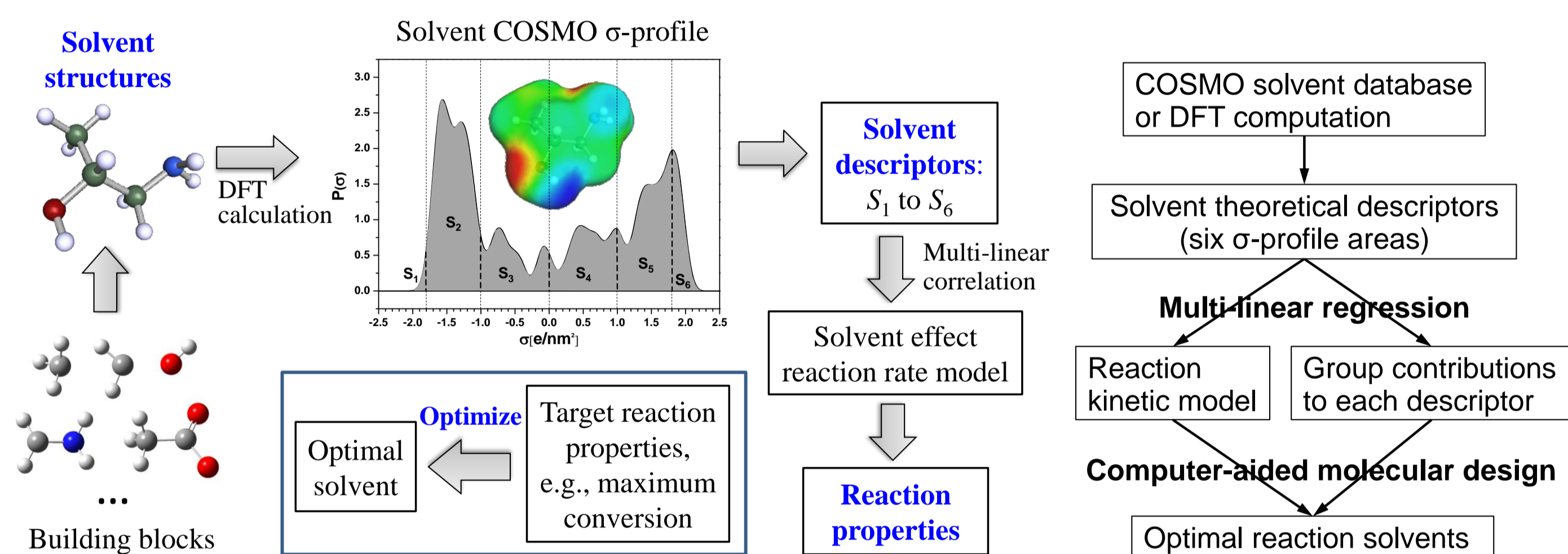
Conductor-like Screening Model (COSMO)

Quantum chemical DFT computation

Screening Charge Density distribution, σ -profile

σ -profile is **molecule-specific** and intensively **related to atom groups**. This inspires us to develop **new solvent theoretical descriptors** to replace conventional solvent parameters in quantifying their effects on chemical reactions.

Methodology and Framework



Reaction solvent design

Computer-Aided Molecular Design: $\max_{n_j (j \in G)} (\log k)$

Rate constant V.S. Groups

$$\log k = \log k_0 + \sum_{i=1}^6 h_i S_i$$

$$S_i = s_{i0} + \sum_{j=1}^N n_j s_{ij}$$

Chemical Feasibility:

$$\sum_{j=1}^N (2 - v_j) n_j - 2m = 0$$

$$n_j (v_j - 1) + 2m - \sum_{k=1}^N n_k \leq 0$$

Chemical Complexity:

$$0 \leq \text{integer } n_j \leq N_{\max}(j)$$

$$\sum_{j \in G_f} n_j \leq 1 \quad \text{if } \sum_{j=1}^N n_j = 2 \quad 2 - \sum_{j=1}^N n_j \leq 0$$

$$\sum_{j \in G_f} n_j \leq 2 \quad \text{else} \quad \sum_{j=1}^N n_j \leq 5$$

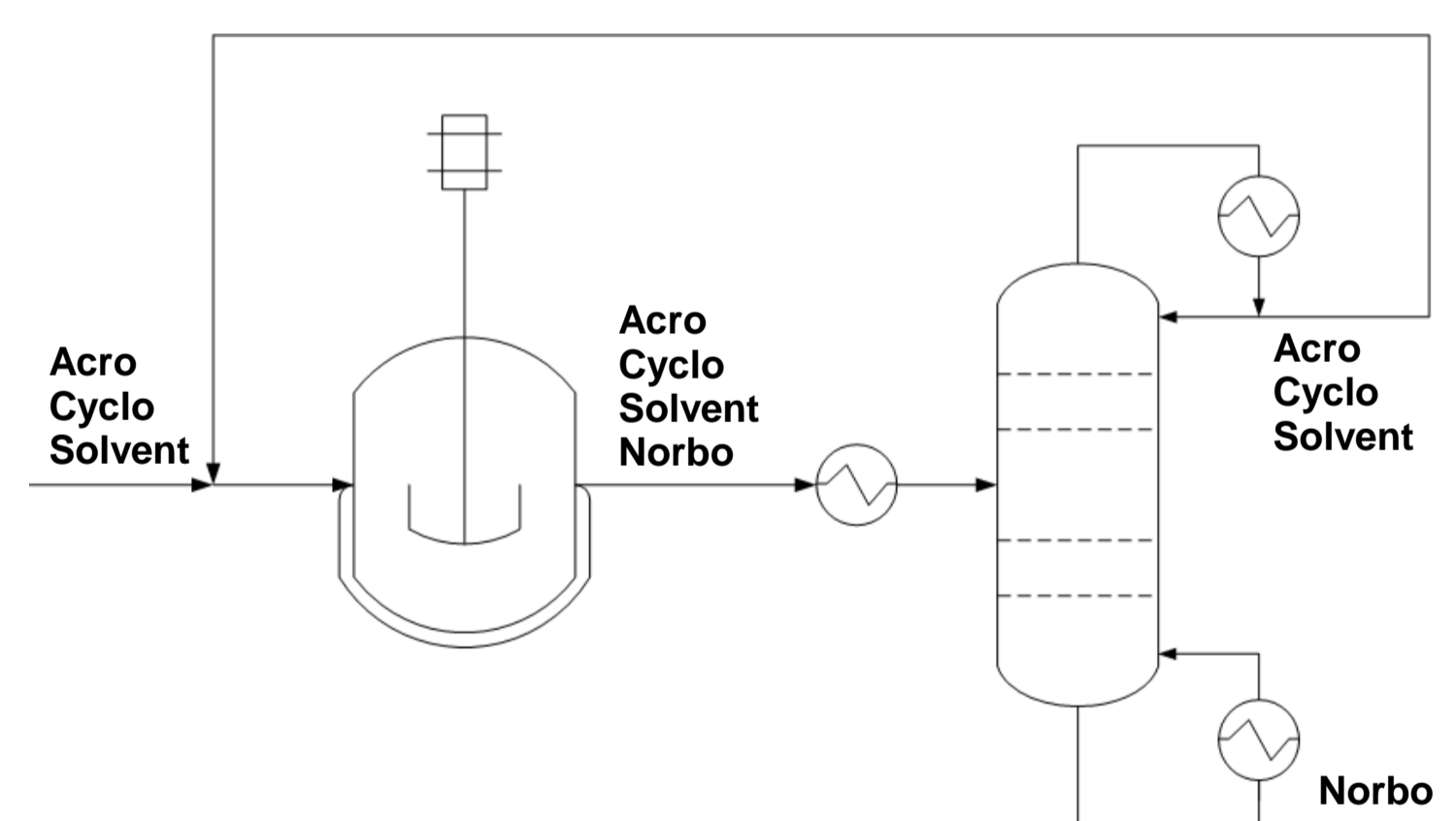
Top 10 solvents from deterministic CAMD optimization:

Group combination	Solvent	$\log k_{\text{pred}}$
2 CH ₃ , 1 C, 1 OH, 1 COOH	(CH ₃) ₂ C(OH)COOH	-1.875
3 CH ₂ , 1 OH, 1 COOH	OHCH ₂ CH ₂ COOH	-1.934
1 CH ₃ , 1 CH ₂ , 1 CH, 1 OH, 1 COOH	CH ₃ CH(OH)CH ₂ COOH	-1.942
2 CH ₂ , 1 OH, 1 COOH	OHCH ₂ CH ₂ COOH	-1.962
1 CH ₃ , 1 CH, 1 OH, 1 COOH	CH ₃ CH(OH)COOH	-1.971
1 CH ₂ , 1 OH, 1 COOH	OHCH ₂ COOH	-1.991
3 CH ₃ , 1 C, 1 COOH	(CH ₃) ₃ CCOOH	-2.406
2 CH ₃ , 1 C, 1 CH ₃ CO, 1 COOH	CH ₃ COC(CH ₃) ₂ COOH	-2.455
2 CH ₃ , 1 C, 1 HCOO, 1 COOH	HCOOC(CH ₃) ₂ COOH	-2.462
1 CH ₃ , 3 CH ₂ , 1 COOH	CH ₃ (CH ₂) ₃ COOH	-2.465

Best solvent found experimentally: acetic acid ($\log k = -2.491$)

Integrated solvent and process design

- ✓ The identification of optimal reaction solvent is far away from the final object of engineers.
- ✓ The integration of solvent design and process optimization based on a process-wide approach is a prerequisite for achieving an optimal chemical process.



- COSMO σ -profile areas (our GC method)

- T_b , T_c , P_c , ΔH_{vap}
- liquid molar volume
- heat capacity

Properties are estimated from different group contribution methods

Reaction solvent design

Reaction rate regression

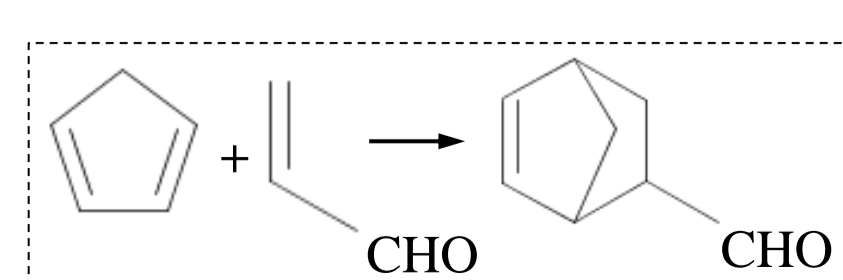
Six σ -profile areas are used to correlate reaction rate constants

$$\log k_{\text{cat}} = \log k_0 + \sum_{i=1}^6 h_i S_i$$

$\log k_0$	-4.0113		
h_1	4.7103	h_4	0.0274
h_2	0.3012	h_5	-0.0292
h_3	0.0050	h_6	3.1087

MAPE = 2.58%
 $R^2 = 0.923$

Solvent	$\log k_{\text{exp}}$	$\log k_{\text{cal}} / \text{Rank}$
Acetic acid	-2.491	-2.492 / 1
Ethanol	-2.964	-3.179 / 3
1-Propanol	-3.186	-3.202 / 4
1-Butanol	-3.219	-3.210 / 5
Methanol	-3.257	-3.080 / 2
Chloroform	-3.383	-3.540 / 7
dichloroethane	-3.602	-3.677 / 9
DMF	-3.640	-3.615 / 8
CH ₂ Cl ₂	-3.699	-3.501 / 6
Toluene	-3.745	-3.786 / 11
Acetonitrile	-3.757	-3.881 / 13
1,4-Dioxane	-3.827	-3.873 / 12
CCl ₄	-3.873	-3.755 / 10
Acetone	-3.983	-3.933 / 14
Ethyl acetate	-4.036	-3.938 / 15



Cyclopentadiene (Cyclo) and Acrolein (Acro) to 5-Norbornane-2-carboxaldehyde (Norbo)

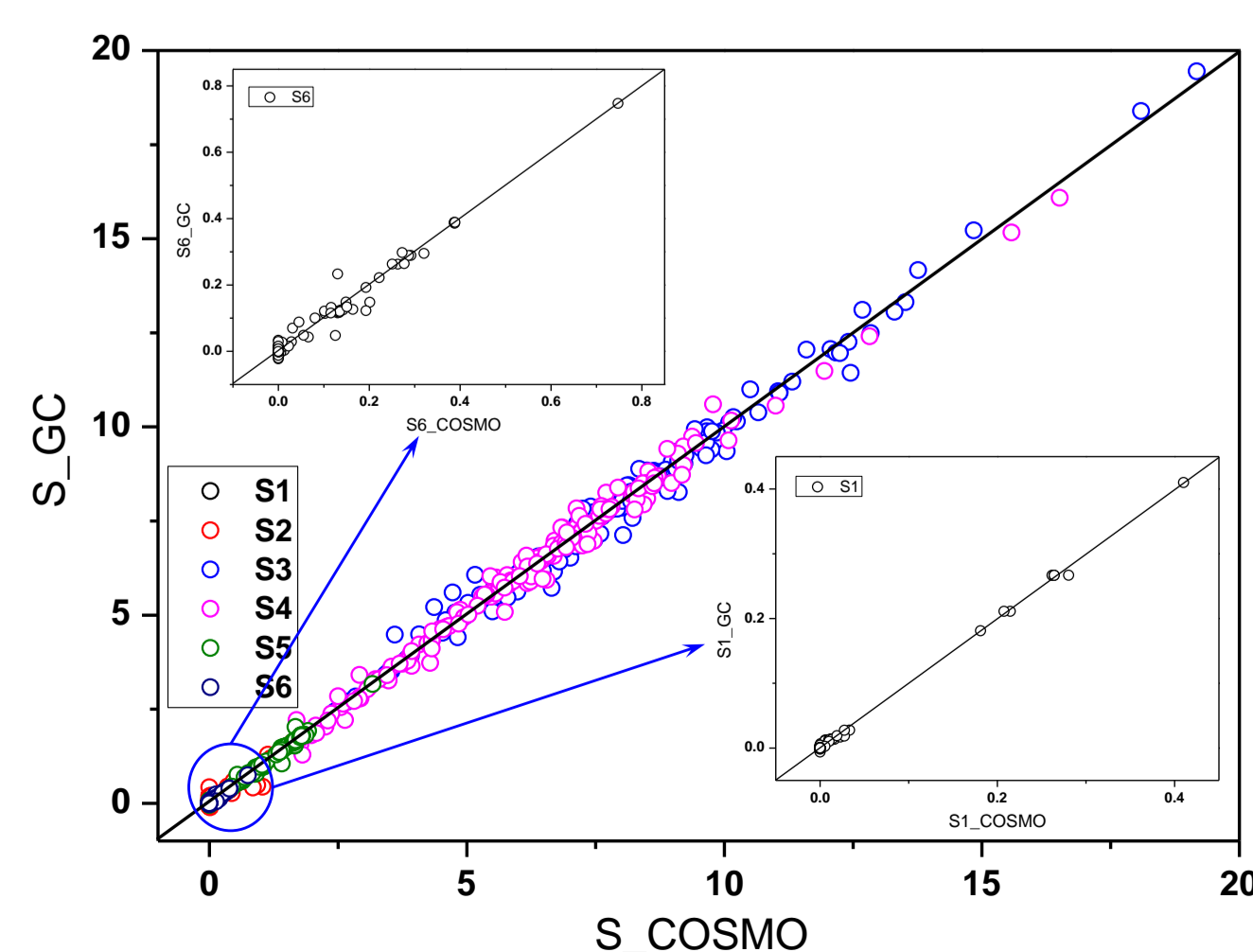
GC method development

Group Contributions (GC) are regressed from σ -profile areas of 168 solvents

$$S_i = s_{i0} + \sum_{j=1}^N n_j s_{ij} \quad (i = 1, 2, \dots, 6)$$

 $(s_{ij}$ is the j -th group's contribution to the i -th area)

	s_{1j}	s_{2j}	...	s_{6j}
CH ₃	0.0009	0.0207	...	-0.0267
CH ₂	0.0000	0.0074	...	-0.0013
CH	-0.0021	0.0122	...	0.0215
C	-0.0044	0.0161	...	0.0736
OH	0.0148	0.7146	...	0.0815
CH ₃ OH	0.0210	0.8769	...	0.1197
H ₂ O	0.0295	1.7598	...	0.3157
CH ₃ CO	0.0012	0.1104	...	-0.0294
CH ₂ CO	0.0006	0.0736	...	-0.0101
CHO	0.0012	0.1153	...	-0.0432
CH ₃ COO	0.0012	0.1153	...	-0.0430
...
COOH	0.2679	0.7073	...	-0.0436



GC predictions on S1 ~ S6 (168 solvents, MAPE = 7.28%)

Acknowledgments

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Publications

- Zhou, T., Qi, Z., Sundmacher, K. Model-based method for the screening of solvents for chemical reactions. *Chemical Engineering Science* 2014, 115: 177-185.
- Zhou, T., McBride, K., Zhang, X., Qi, Z., Sundmacher, K. Integrated solvent and process design exemplified for a Diels-Alder reaction. *AIChE J.* 2014, doi: 10.1002/aic.14630.
- Zhou, T., Sundmacher, K. Robust design of optimal solvents for chemical reactions - a combined experimental and computer-aided strategy (to be submitted).