

Modelling hygroscopicity-induced gas-aerosol partitioning, organic surface enrichment and cloud droplet formation

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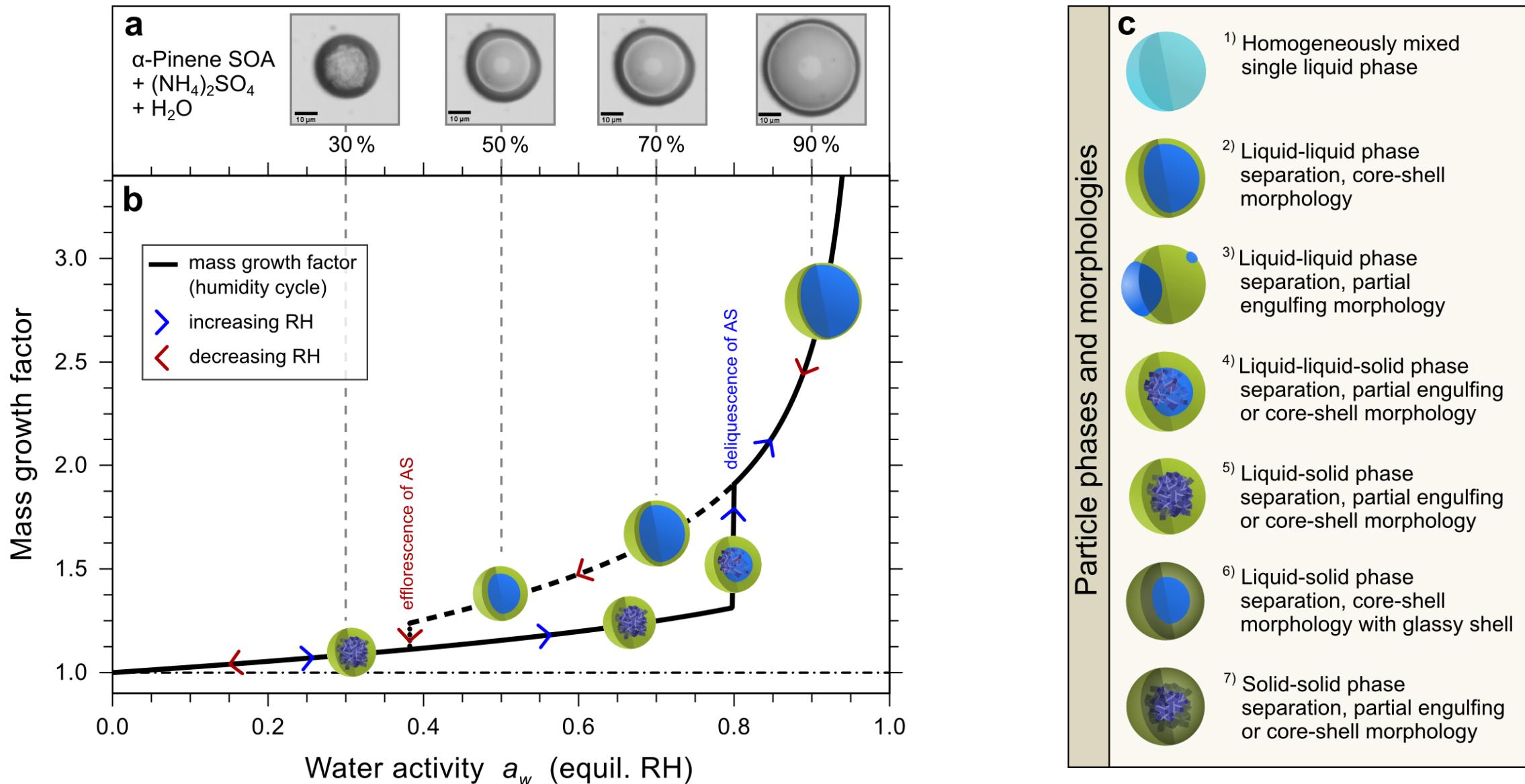


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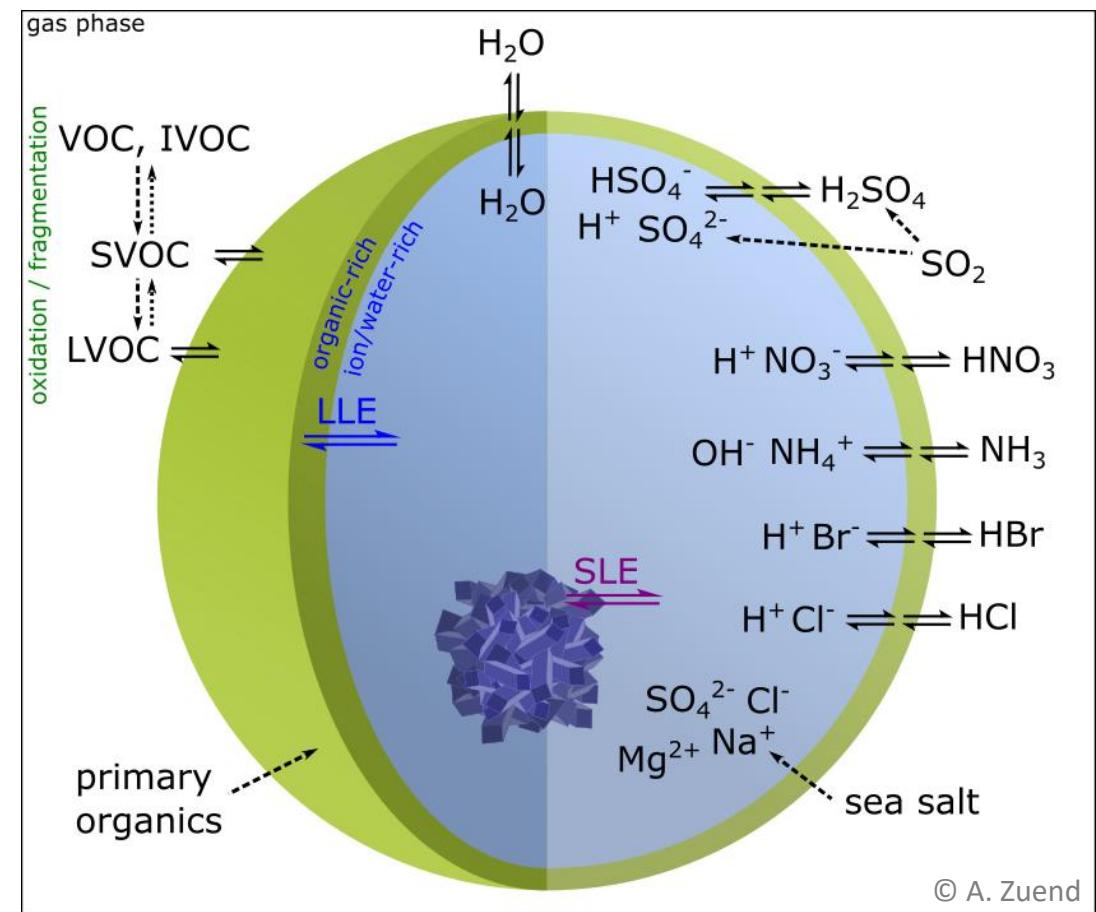
Introduction: organics & water-driven phase transitions



Organic–inorganic aerosol thermodynamics

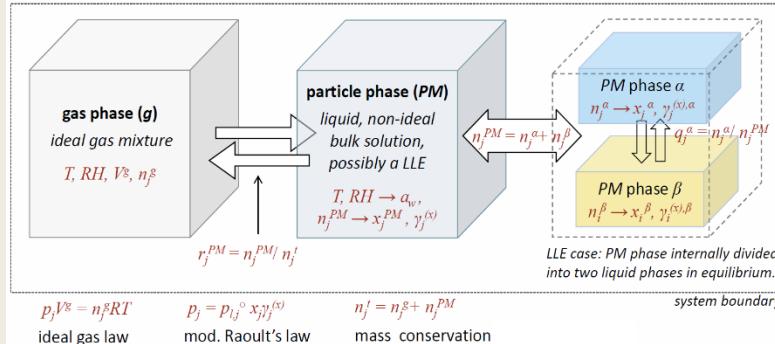
AIOMFAC-based modeling framework

- Gas–particle partitioning at different RH
- Aerosol phases, physical states, viscosity
- Non-ideal mixing, liquid–liquid equilibrium (LLE)
- Aerosol acidity (multiphase; e.g. Pye et al, 2020, *ACP*)
- Hygroscopicity & cloud condensation nucleus (CCN) properties



Modeling non-ideal mixing thermodynamics & gas-particle partitioning

Gas-particle equilibrium model:



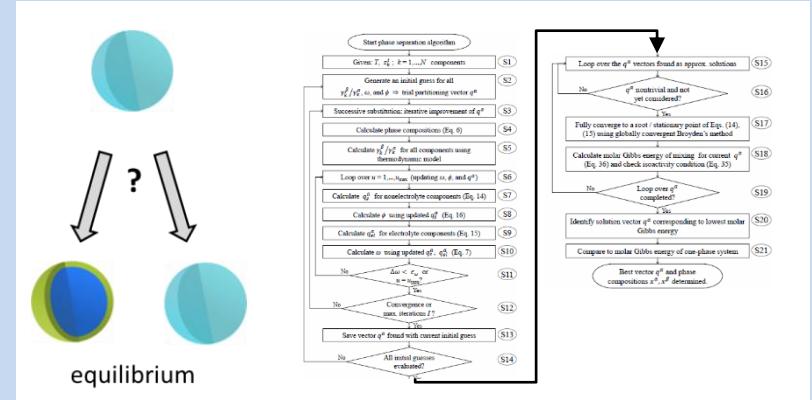
[e.g., Zuehd et al., 2010, ACP]

Volatility: Pure component vapour pressures as function of T (measured or predicted)

Output: (for specified input system)

- # liquid phases;
(& optionally solid phases)
- chemical compositions of all phases
(bulk prediction)

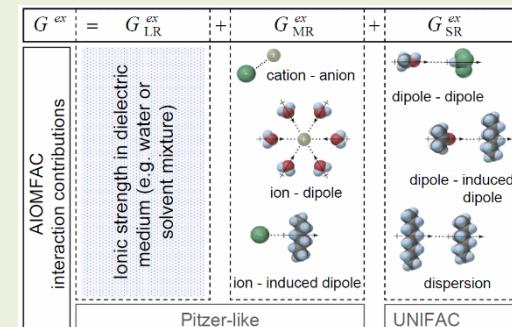
LLE algorithm: Liquid–liquid phase separation?



[e.g., Zuehd and Seinfeld, 2013, Fluid Phase Equilib.]

AIOMFAC model:

Organic—inorganic mixing in a liquid phase



[Zuehd et al., 2008; 2011, ACP; <https://aiomfac.lab.mcgill.ca>]

Next 3 slides outline examples of applications
of the AIOMFAC model and the AIOMFAC-based
equilibrium framework

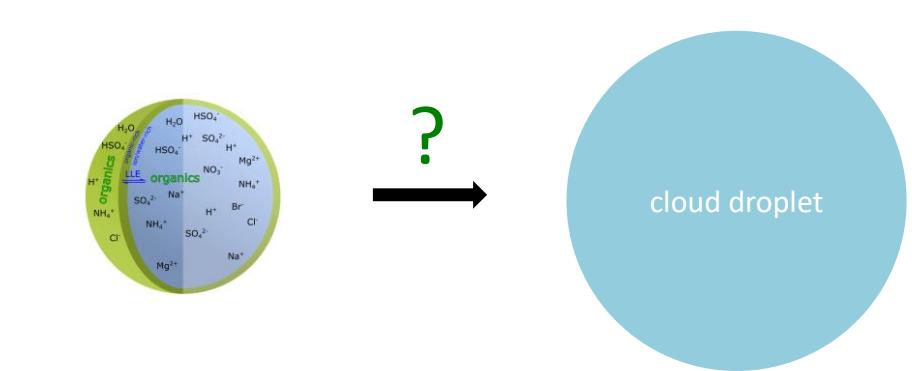
Application of AIOMFAC framework: CCN activation & the role of organics

Köhler theory:

$$S = a_w \exp \left[\frac{4 \sigma M_w}{RT \rho_w D_0 \text{ HGF}} \right] \quad \text{with} \quad \text{HGF} = \frac{D}{D_0}$$

What is the role of organic aerosol constituents & liquid–liquid phase separation in affecting CCN properties?

- Thermodynamic perspective
- Aerosol population / air parcel dynamics



Field study at Mace Head: effective ultrafine CCN in marine air

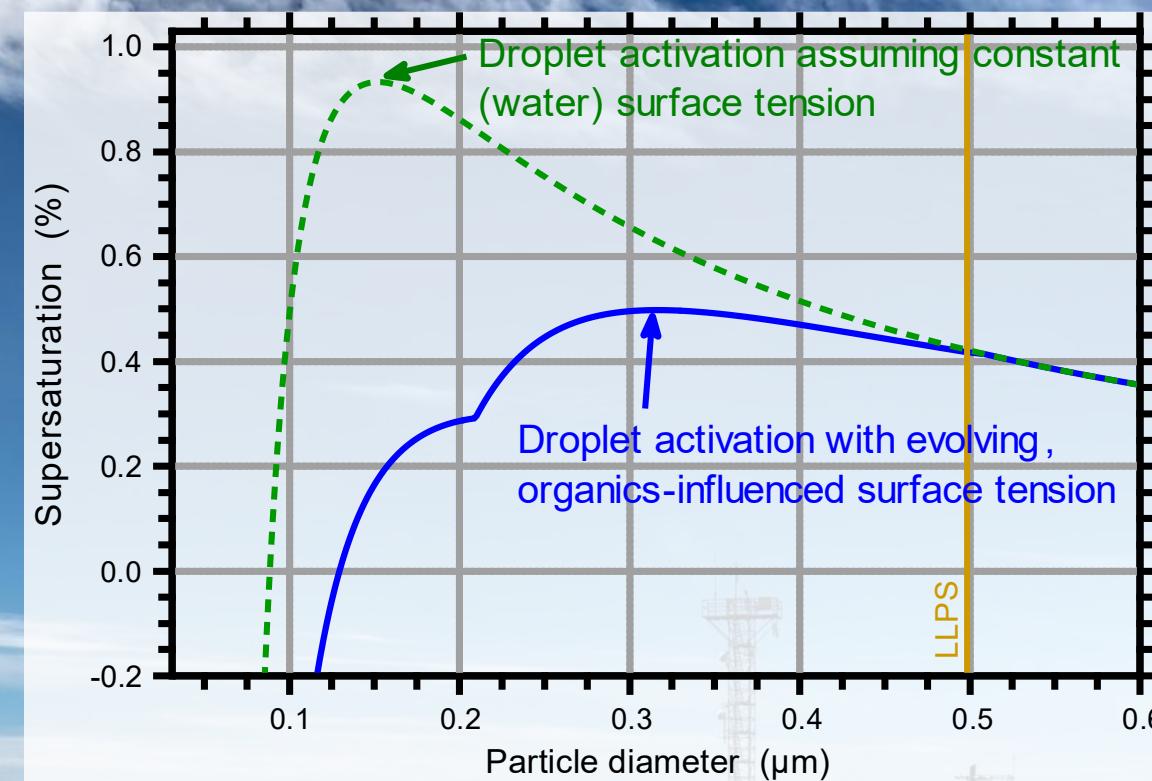
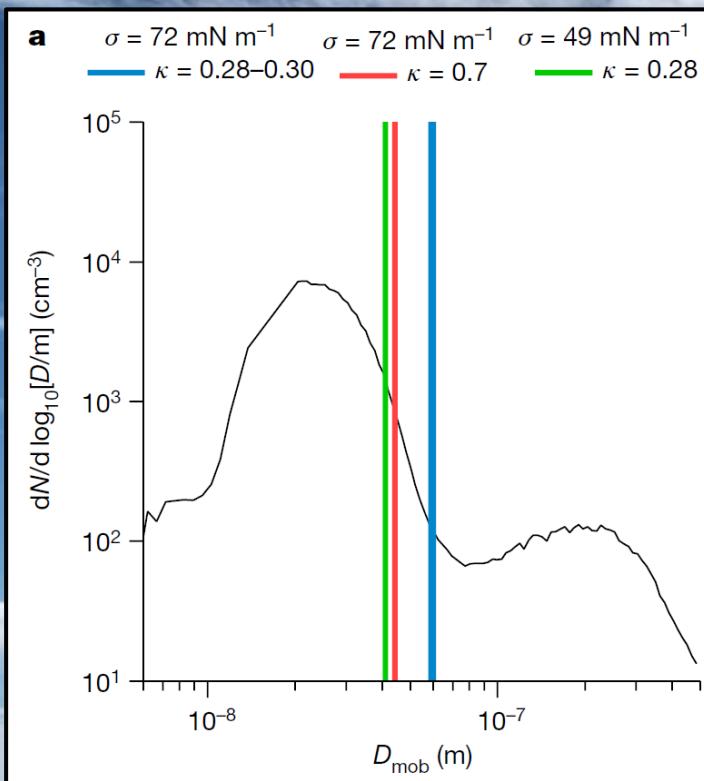
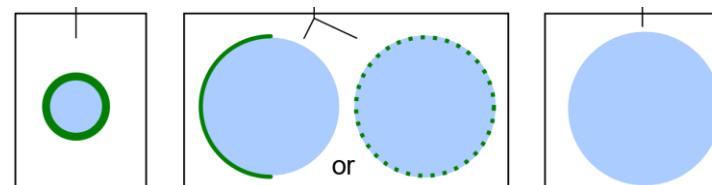


Photo: Mace Head Atmospheric Research Station, C-CAPS NUI Galway, Ireland

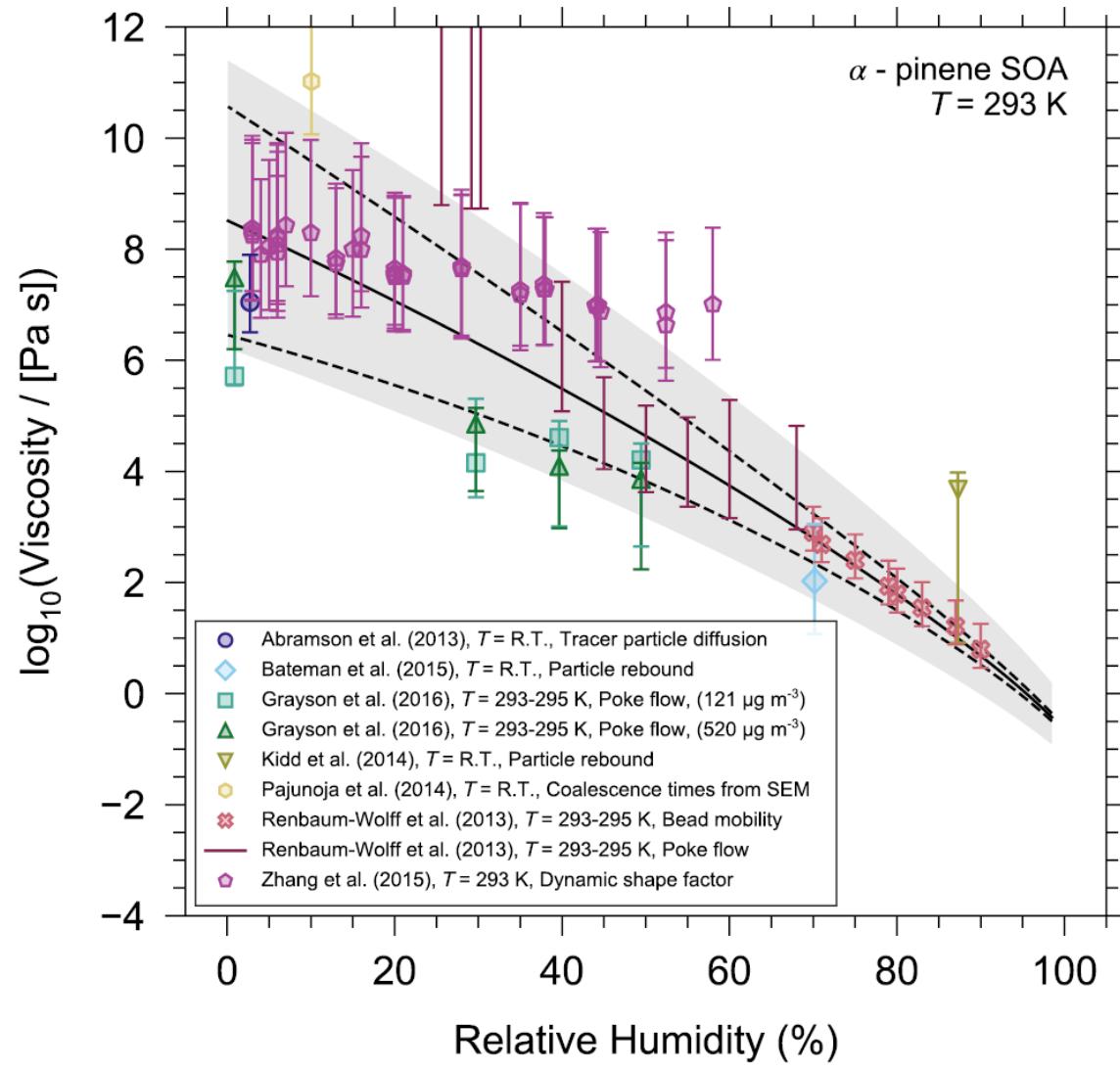
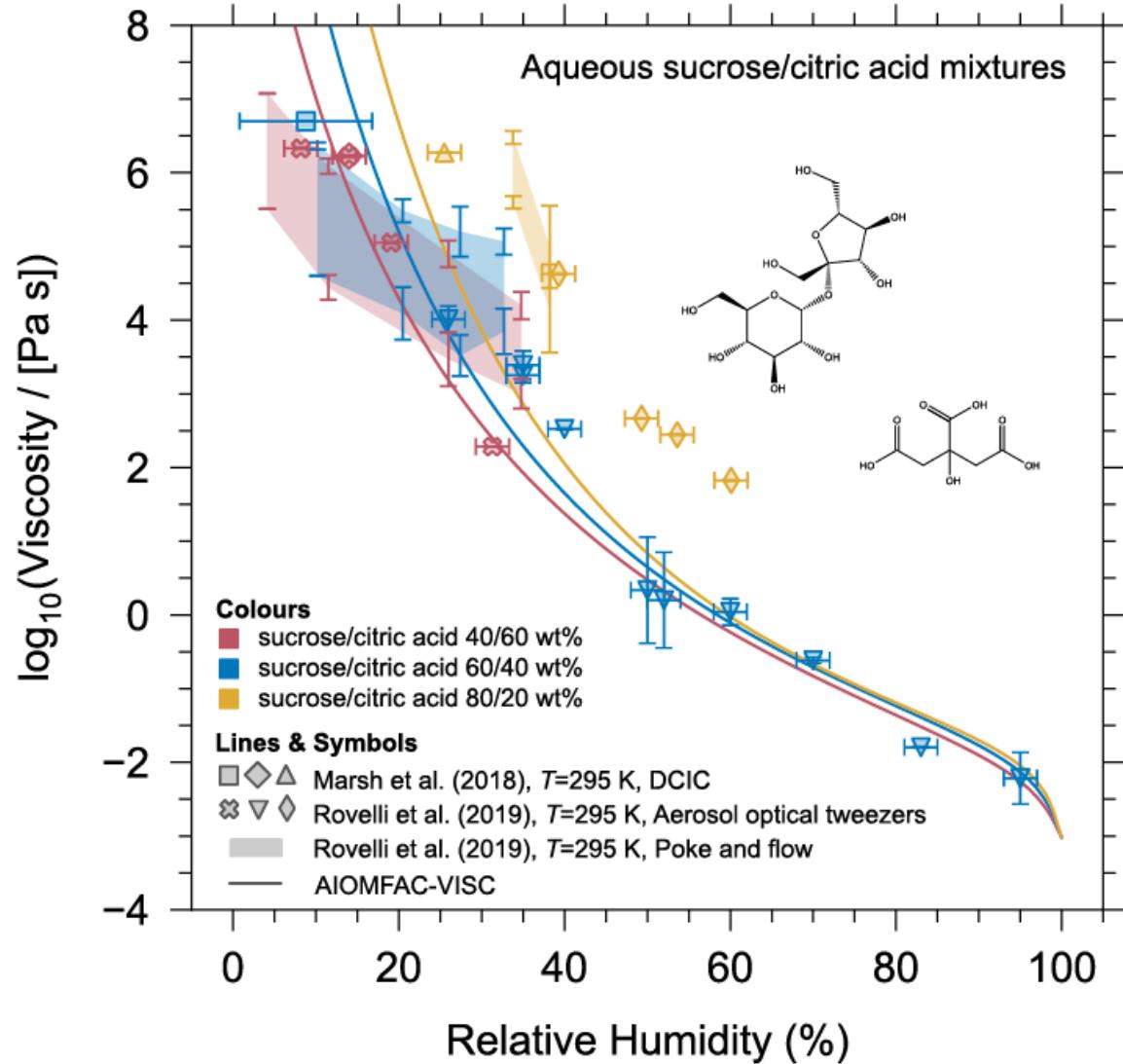
- **Proposed explanation:**

evolution of surface tension due to organic-rich shell phase (surface tension lowering), depending on surface coverage;



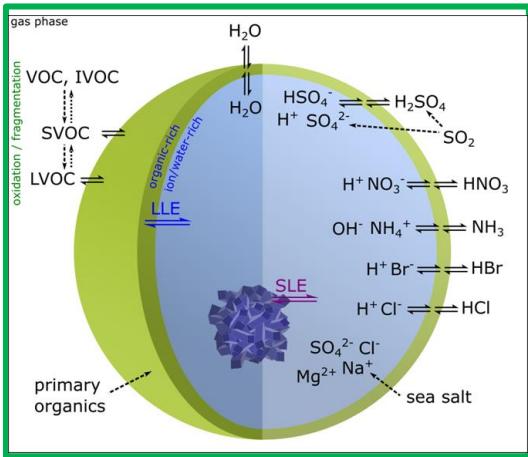
[Ovadnevaite et al., 2017, *Nature*] 7

AIOMFAC-VISC: Predictive aqueous organic aerosol viscosity

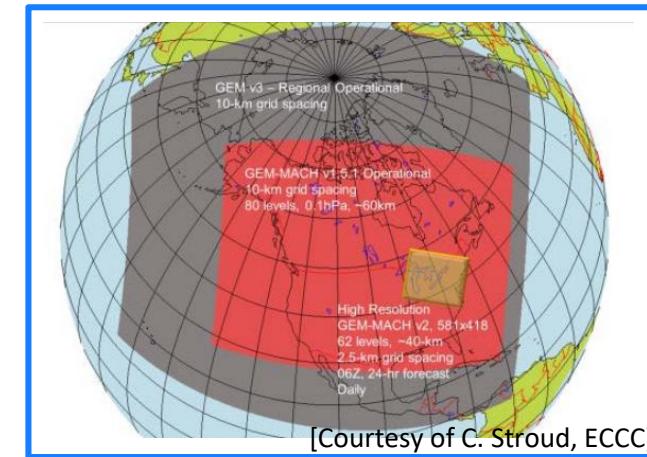


From process level to 3-D models

Process level



3-D atmospheric models



Chemical-structure-based models

- for detailed system characterization
- computationally expensive

Reduced-complexity models

- processing of field measurement data
- operating on limited information



Equilibrium gas-particle partitioning methods

Volatility-based

can use commonly measured (bulk) properties.



Building an intermediate, reduced-complexity model

Volatility-based

can use commonly measured (bulk) properties.

2-product

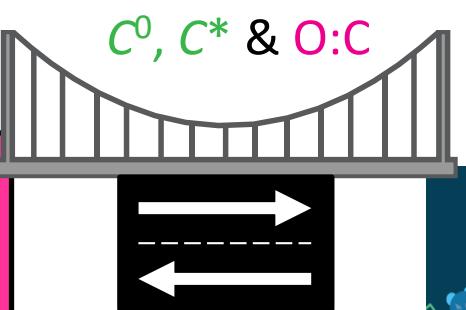
2D-VBS

VBS

(volatility bins)

C^* , (K_p) O:C

no water content considered
assumes $\gamma \approx 1$ (ideal mixing)



water-sensitive



Structure-based models

organic Inorganic

AIOMFAC

E-AIM

UNIFAC

ISORROPIA

molecular structure inputs,
water considered

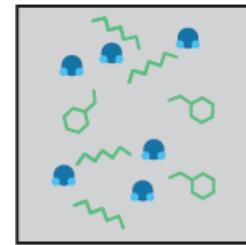
non-ideal liquid mixing

Binary Activity Thermodynamics (BAT) model



- **Organic \leftrightarrow water interactions**
- Generalized Duhem–Margules activity coefficient model

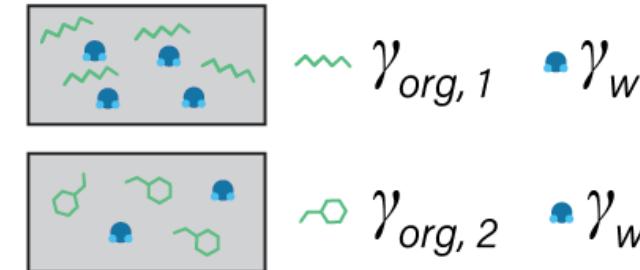
(a) Multi-component mixture



$\gamma_{org, 1}$
 $\gamma_{org, 2}$
 γ_w

Non-ideal interactions
among all species

(b) BAT approximation



Non-ideal interactions
with water

BAT model – a general parameterization

Generalized Duhem–Margules binary activity coefficient model of BAT:

$$g^e/RT = C_1 \varphi_{org}^* (1 - \varphi_{org}^*) + C_2 \varphi_{org}^* (1 - \varphi_{org}^*)(1 - 2\varphi_{org}^*)$$

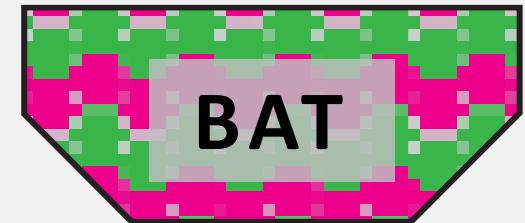
φ_{org}^* = scaled volume fraction of organic

$$C_1 = a_1 \exp(a_2 \cdot O:C) + a_3 \exp\left(a_4 \frac{M_{water}}{M_{org}}\right)$$

$$C_2 = a_5 \exp(a_6 \cdot O:C) + a_7 \exp\left(a_8 \frac{M_{water}}{M_{org}}\right)$$

→ Fit parameters a_1, a_2, \dots for a wide class of organics

Input
O:C, (H:C)
 M_{org}
mole fraction of org./water

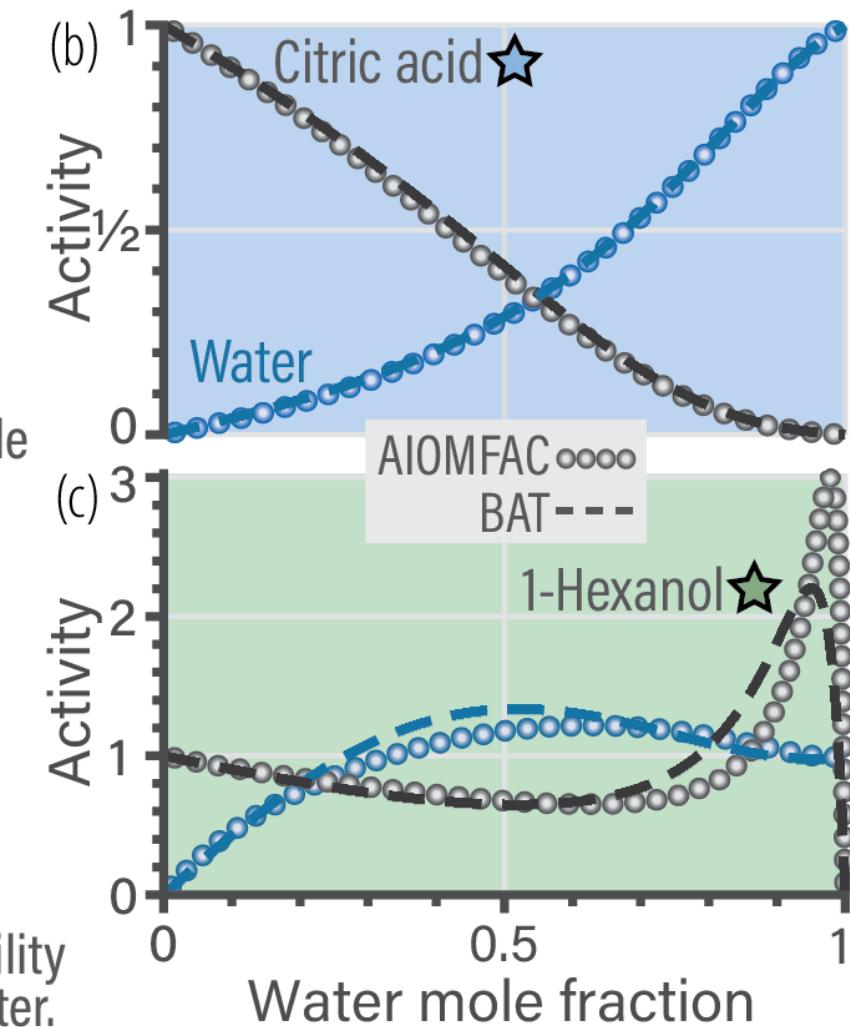
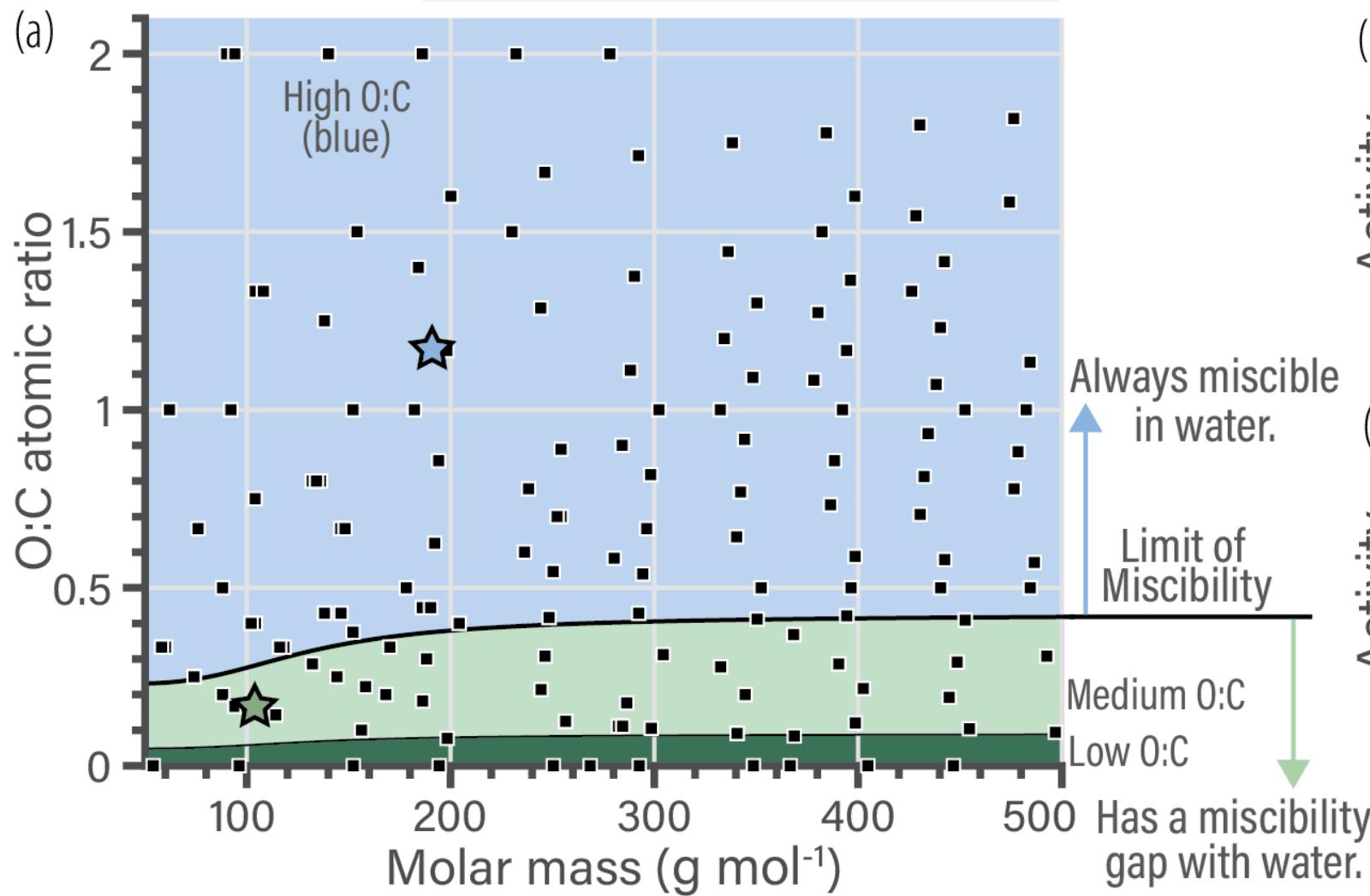


Output
 $\gamma_{org}, a_{org},$
 $\gamma_{water}, a_{water}$ (equil. RH)

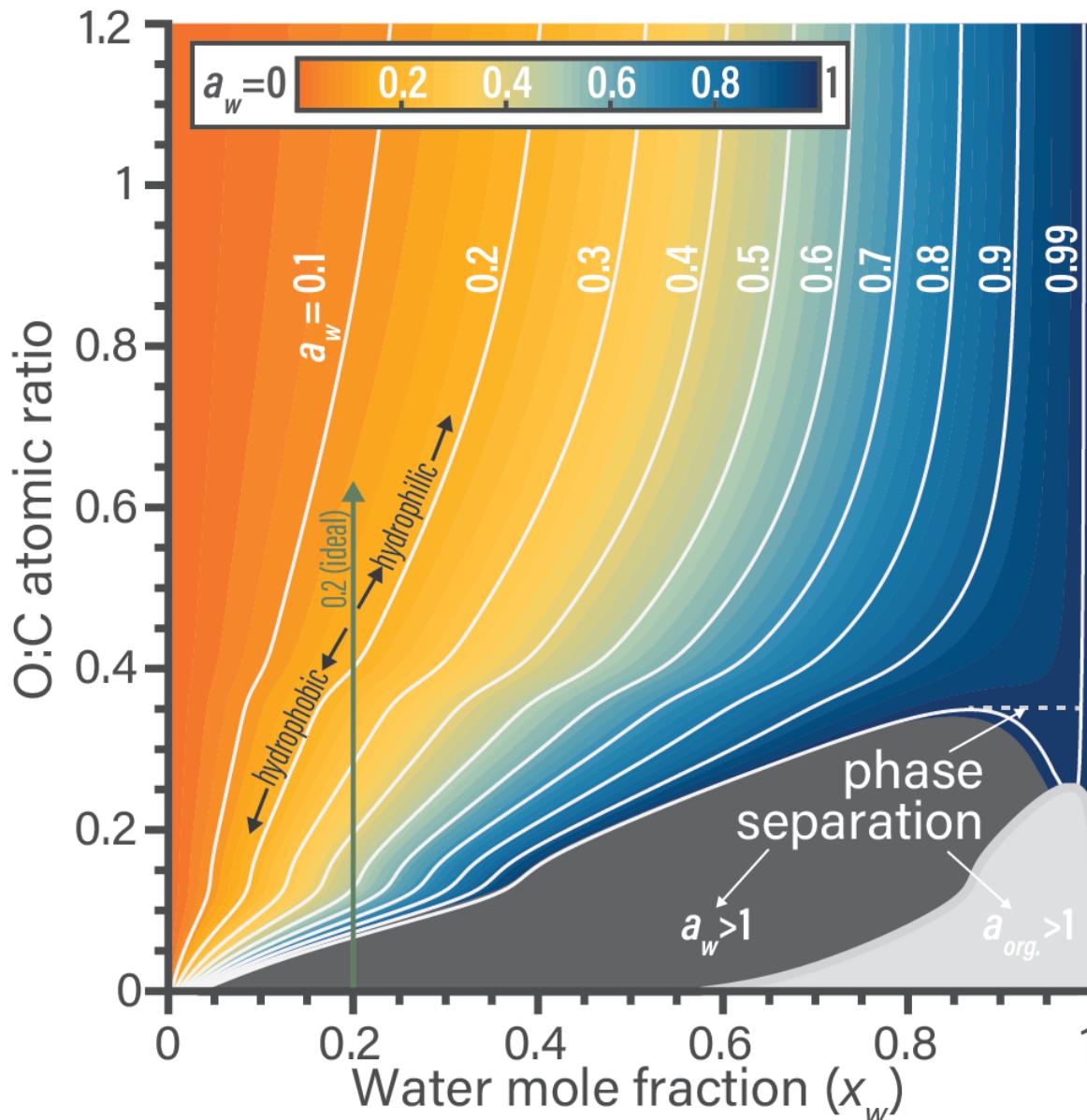
Training a **BAT** – 3 distinct domains



Training dataset of 160 molecules



BAT predictions for non-ideal mixing behavior



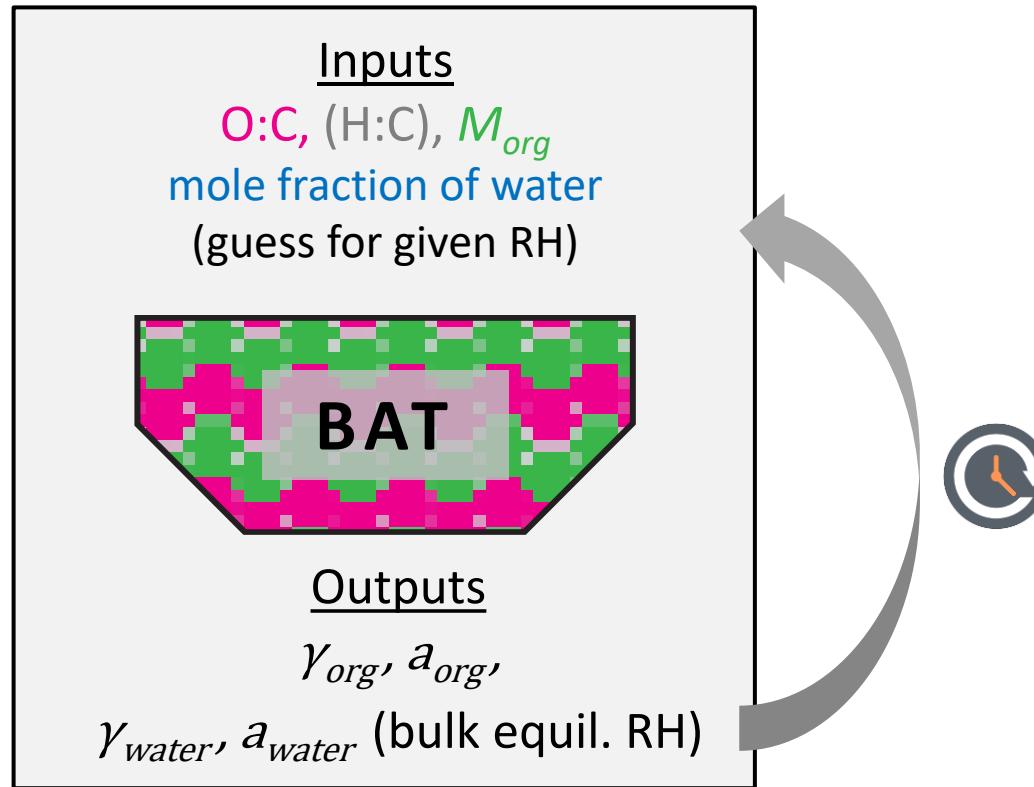
BAT prediction of water activity
Binary Activity Thermodynamics

- here $M_{org.} = 200$ g/mol → scan O:C, x_w
- $a_w = x_w \times \gamma_w$ (= bulk equil. RH)
- BAT reveals range of phase separation via a simplified LLPS estimation method (for high efficiency)

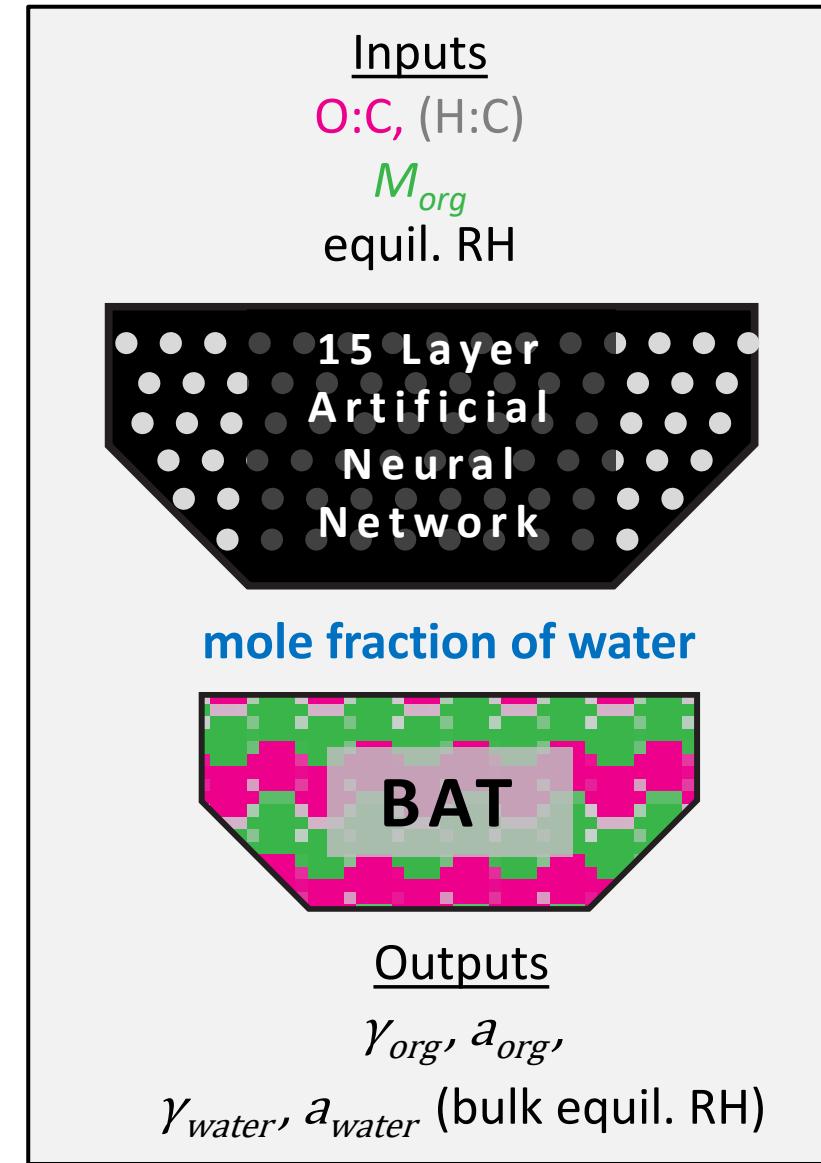
Organic water uptake: iteration vs. neural network

In SOA experiments and chemical transport models the **RH**, **O:C**, **(H:C)** and **M_{org}** are typically provided.

Iteration: find aerosol water content **for given RH**

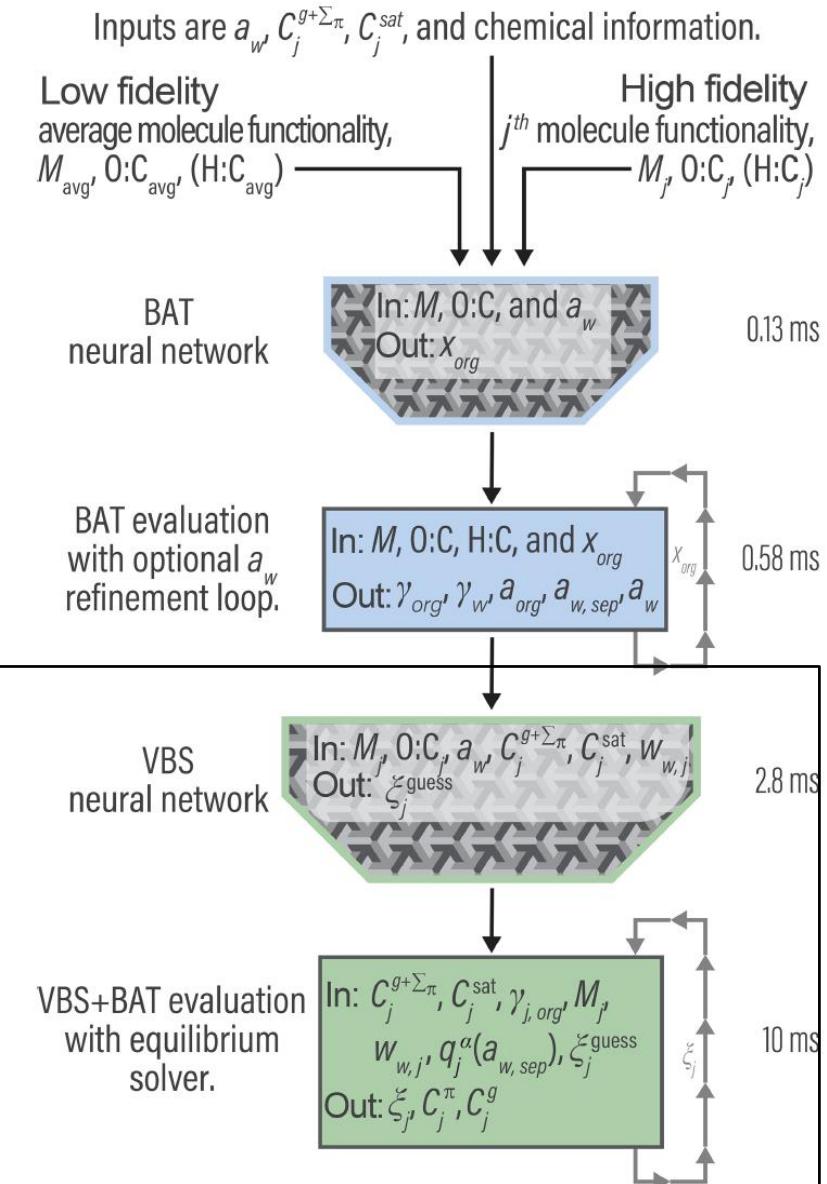
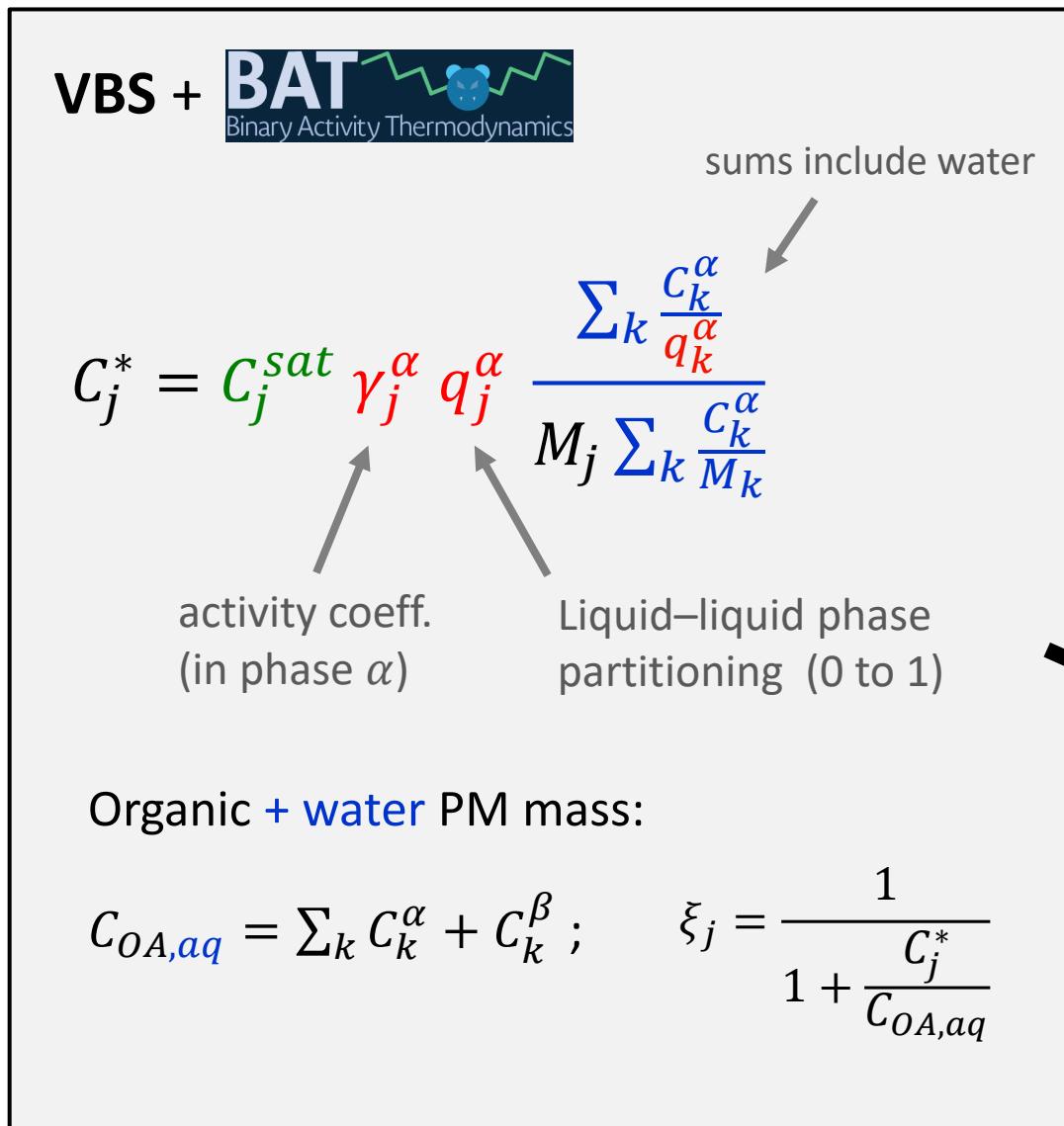


iterative → time consuming

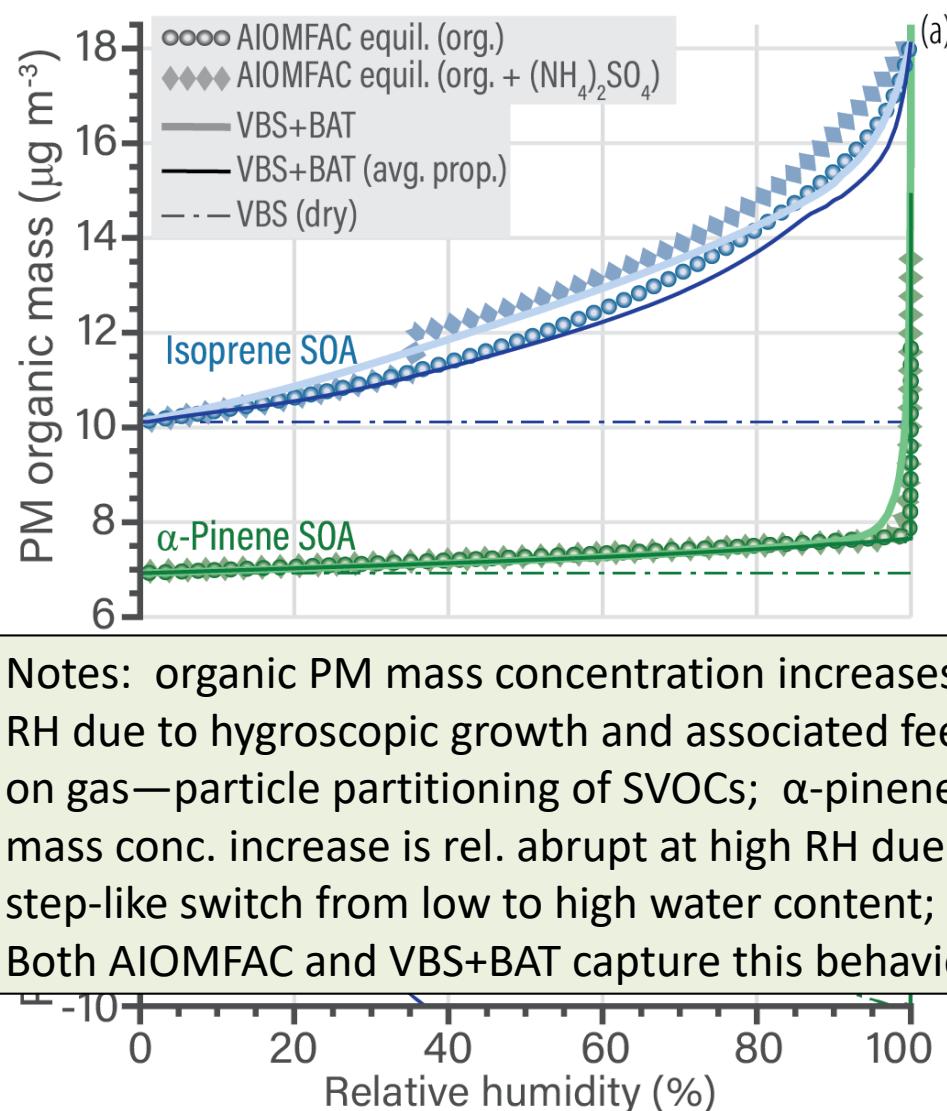


with trained NN, non-iterative → substantially faster

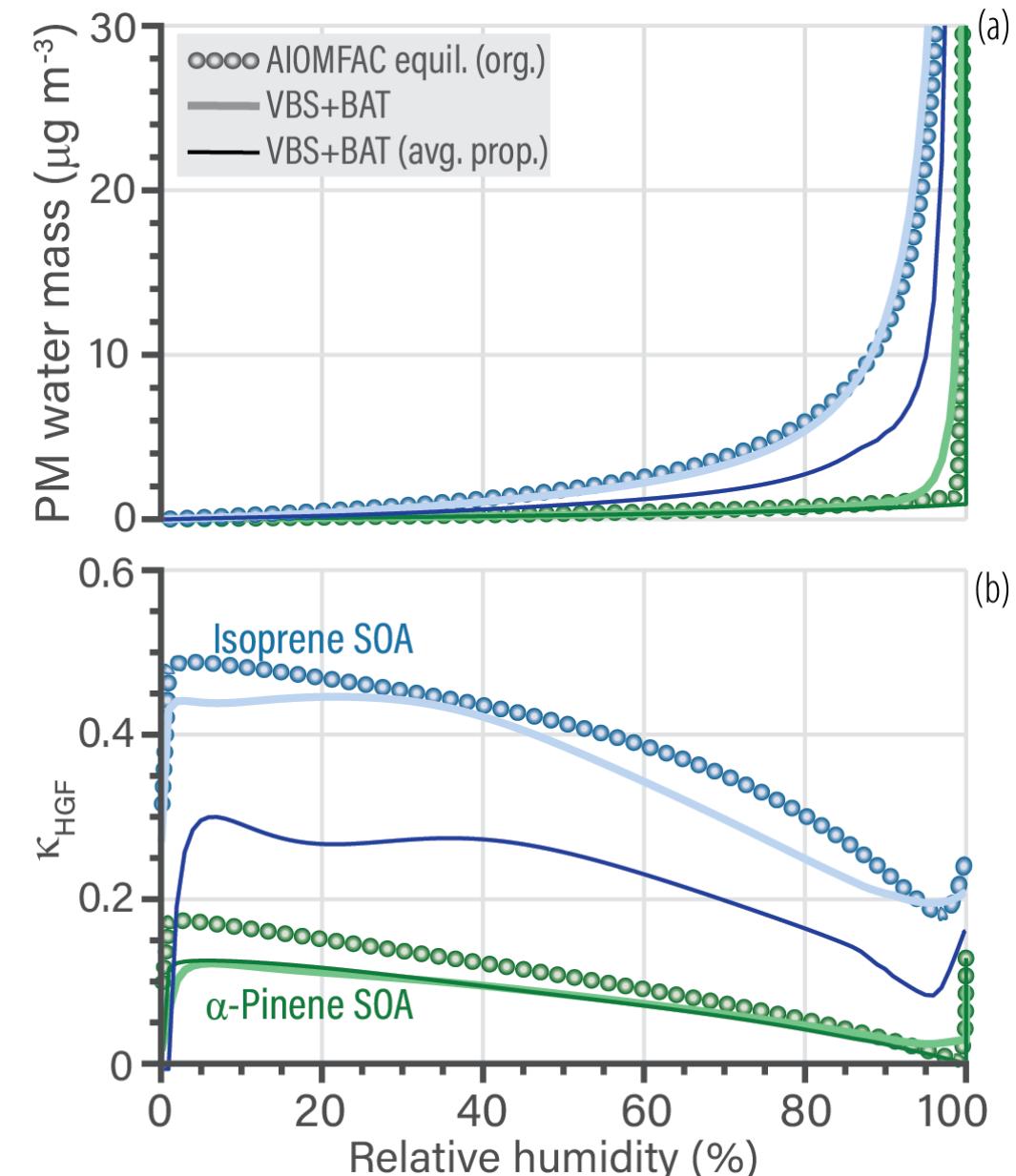
VBS + BAT: including water, γ_j , and phase separation



VBS + BAT vs. AIOMFAC predictions: RH-dependent aerosol mass



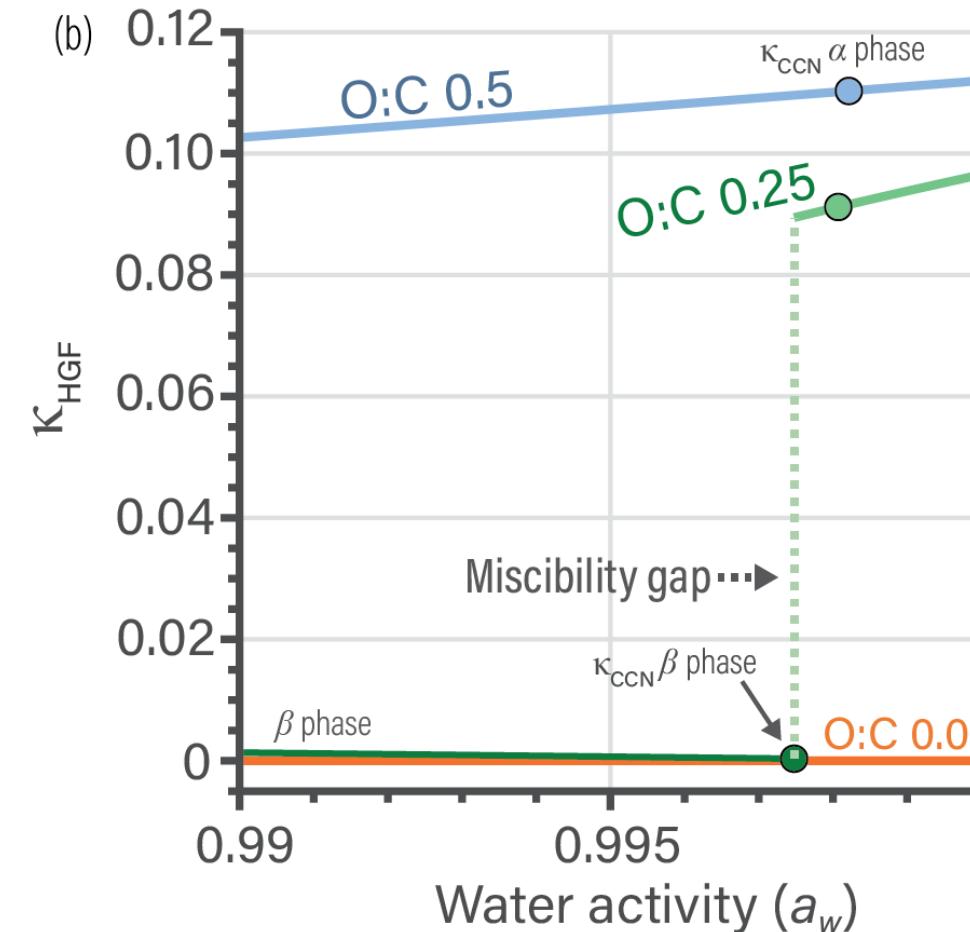
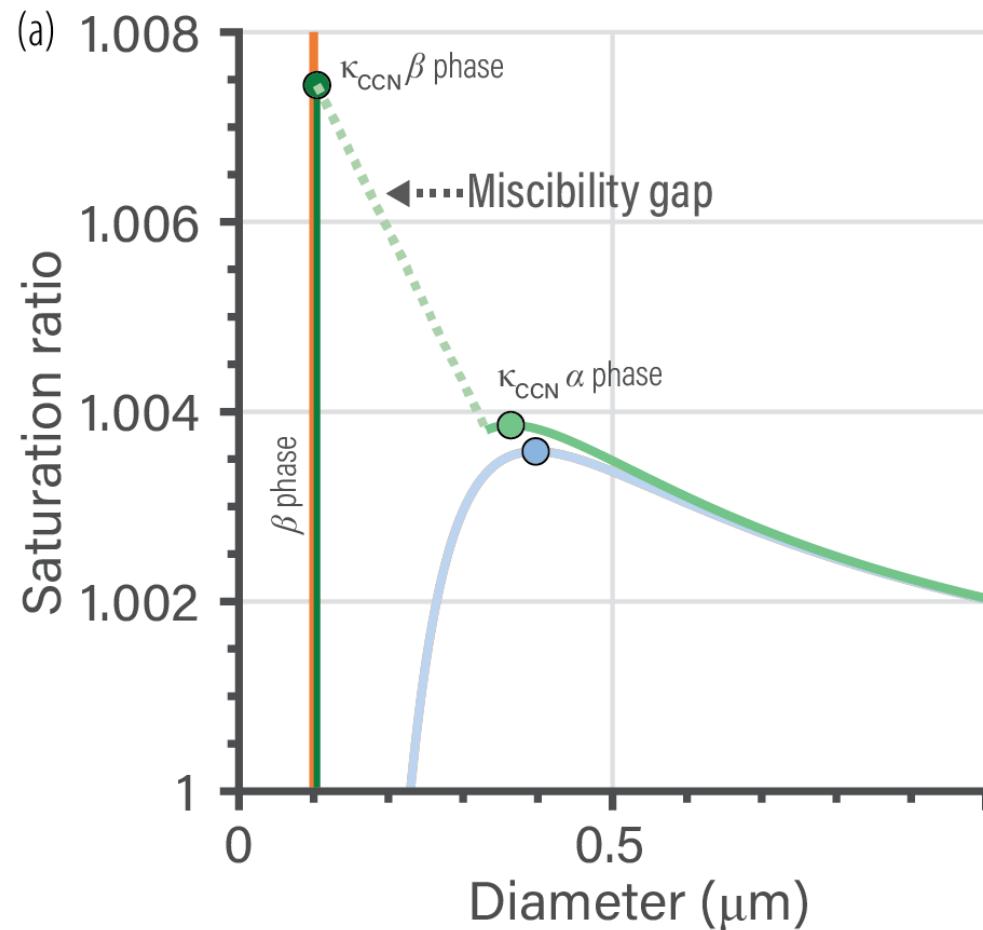
[Gorkowski et al., 2019, ACP]



BAT predictions: O:C-dependent CCN activation

Particle of 100 nm dry diameter and $M_{\text{org}} = 200 \text{ g/mol}$

[Gorkowski et al., 2019, ACP]

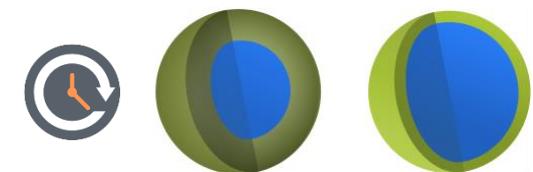
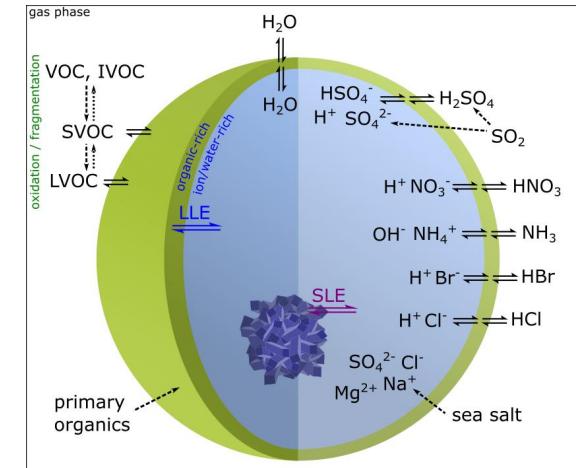


Notes: BAT model allows for predictions of Köhler curves (left) and associated effective kappa values (right) in the context of scanning the O:C / M_{org} parameter space;
→ High-RH behavior and CCN activation point can be predicted for (generic) organic aerosols.

Summary

Prediction of water uptake and related effects on aerosol and CCN properties with different tools:

- AIOMFAC & AIOMFAC-based thermodynamic equilibrium framework
- Organic aerosol viscosity prediction (AIOMFAC-VISC)
- Binary Activity Thermodynamics (BAT) reduced-complexity model
- VBS + BAT
 - organic PM water content (hygroscopicity)
 - organic co-condensation effects with increasing RH
 - can account for liquid–liquid phase separation.



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References



- **Gorkowski, Preston & Zuend (2019):** Relative-humidity-dependent organic aerosol thermodynamics via an efficient reduced-complexity model, *Atmos. Chem. Phys.*, 19, 13383–13407, <https://doi.org/10.5194/acp-19-13383-2019>, 2019.

AIOMFAC-VISC

- **Gervasi, Topping & Zuend (2020):** A predictive group-contribution model for the viscosity of aqueous organic aerosol, *Atmos. Chem. Phys.*, 20, 2987–3008, <https://doi.org/10.5194/acp-20-2987-2020>, 2020.
- <https://aiomfac.lab.mcgill.ca>

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