

# Rapid Collection of Experimental Physicochemical Property Data to Inform Various Models and Testing Methods

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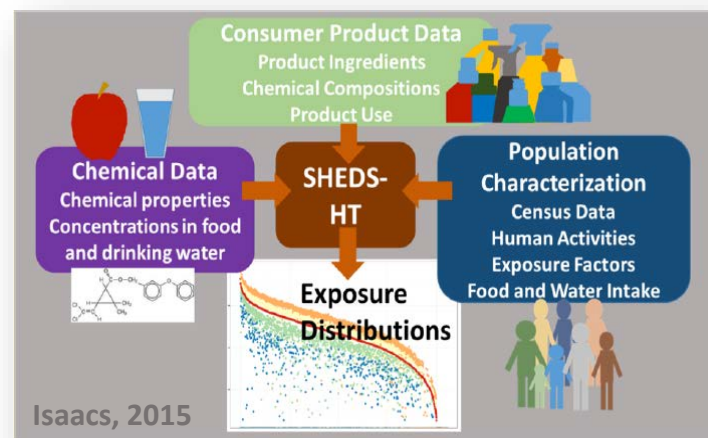
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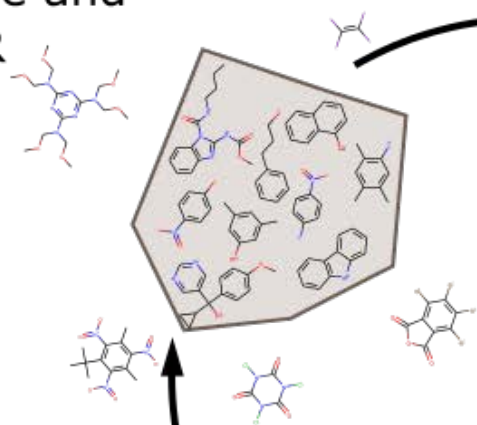
# The Need for Reliable Physicochemical Properties

- High-throughput models have been developed to estimate exposure and dose of substances from near-field sources
- Physicochemical properties parameterize many of these models
- Quantitative structure property relationship (QSPR) models to predict these properties have been developed to fill data gaps
- Most measured datasets of “biologically relevant” physicochemical properties are for substances relevant to the pharmaceutical industry

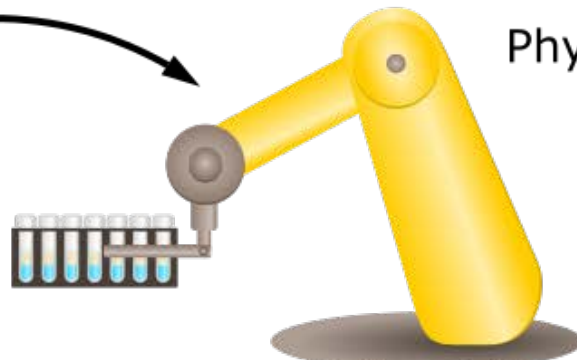


# Continuous Updating of QSPR Models

Select New Chemicals for Estimation Inside and Outside of QSPR Applicability Domains



Perform Experimental Estimates of Physicochemical Properties



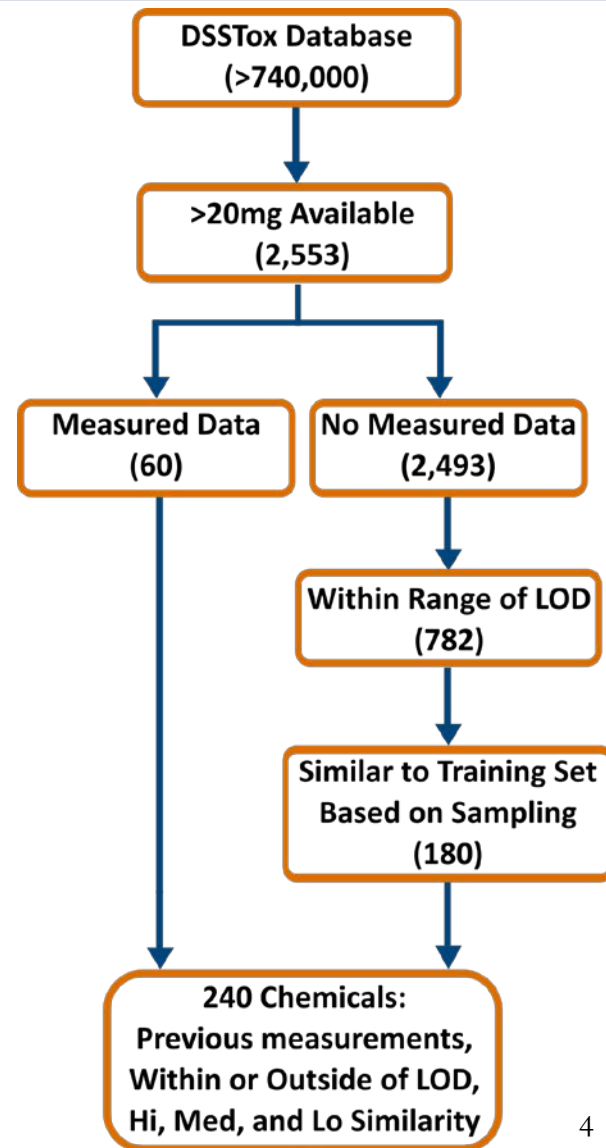
Evaluate Estimates and Inform Models



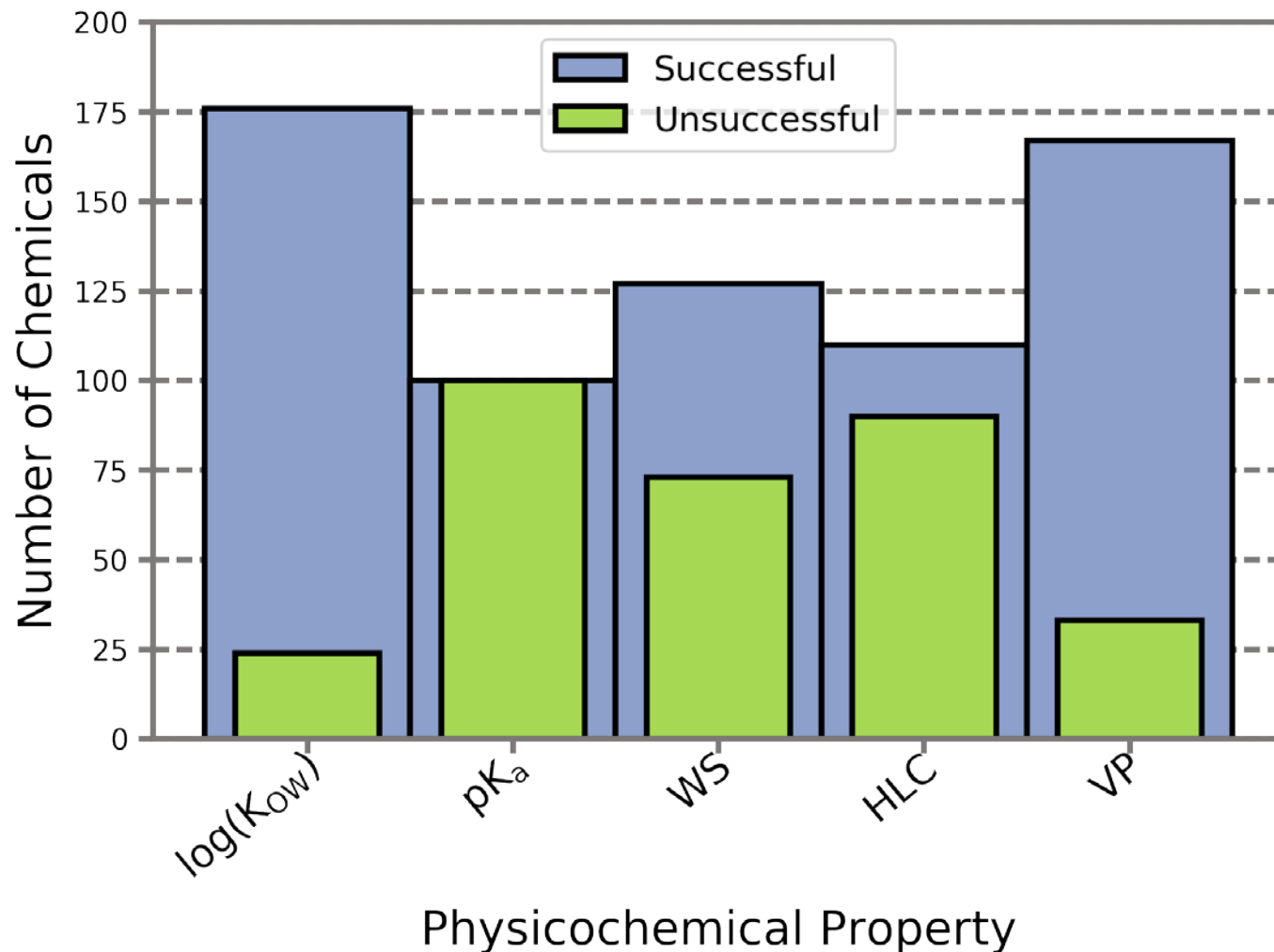
- New measured data are needed for improving QSAR applicability domains

# Process for Selecting Pilot Chemicals for Testing

- If we select the same chemicals in current training set (PHYSPROP), then we get a nice comparison, but we get no expansion of chemical space
- If we select all new chemicals, then we get no comparison, but an expansion of chemical space
- For the pilot, 200 chemicals were submitted for measurement of 5 properties:
  - $\text{Log}_{10}K_{ow}$
  - Henry's Law constant (HLC)
  - Vapor pressure (VP)
  - Water solubility (WS)
  - Acid dissociation constant ( $\text{pK}_a$ )

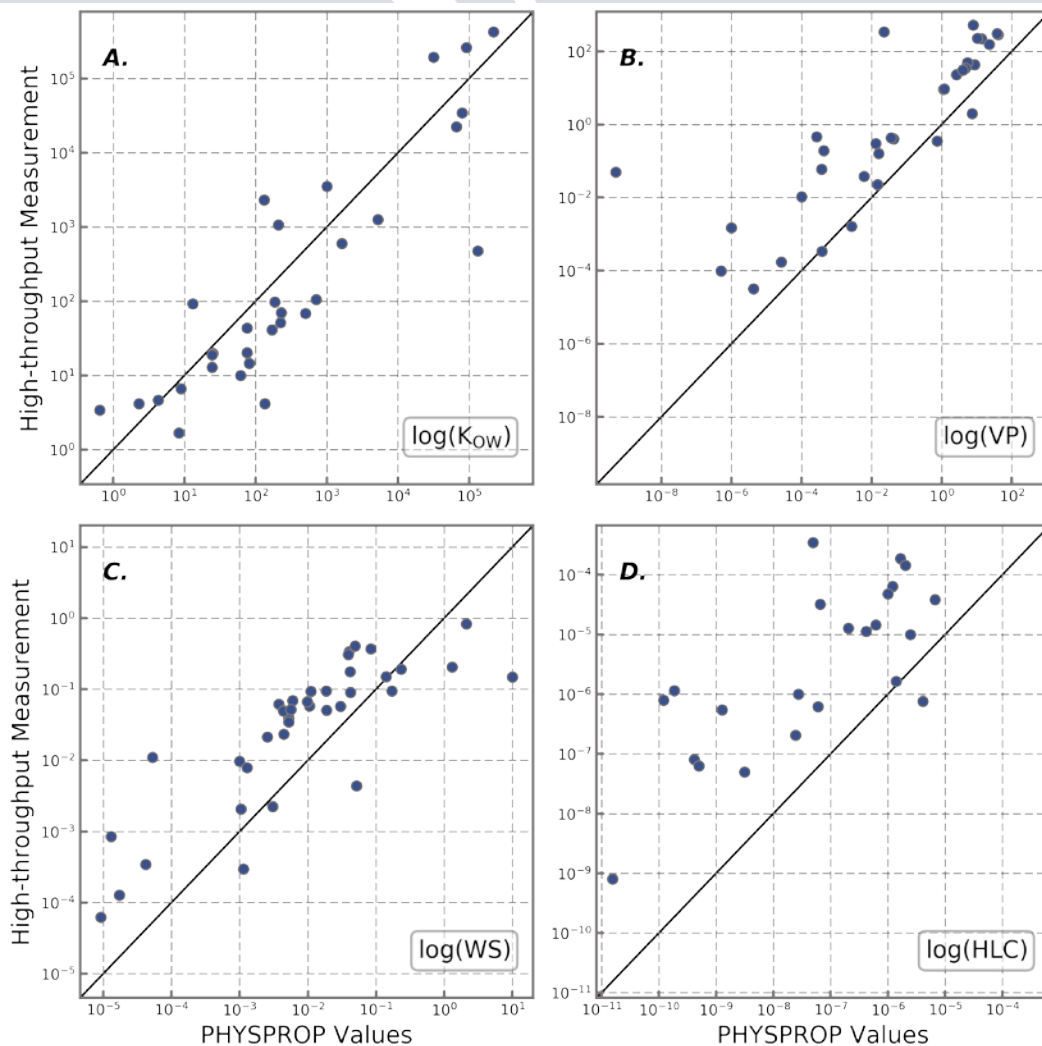


# Summary of Successful Measurements



- Log<sub>10</sub>K<sub>ow</sub> returned the greatest number of measurements and WS, the least

# Success for Compounds in Existing Training Set



- Correspondence between new measurements and those previously reported

# QSPR Tools and the Property Predictions Used

- ACD/Labs
  - Advanced Chemistry Development, Inc.
  - $\text{Log}_{10}K_{\text{ow}}$ , WS, and  $\text{pK}_{\text{a}}$
- ChemProp
  - Chemical Properties Estimation Software System (UFZ)
  - $\text{Log}_{10}K_{\text{ow}}$ , VP, and WS
- EPISuite
  - Estimation Program Interface
  - $\text{Log}_{10}K_{\text{ow}}$ , VP, WS, and HLC
- NICEATM
  - NTP Interagency Center for the Evaluation of Alternative Toxicological Methods
  - $\text{Log}_{10}K_{\text{ow}}$ , and WS
- OPERA
  - OPEn (quantitative) Structure-activity Relationship Application
  - $\text{Log}_{10}K_{\text{ow}}$ , VP, WS, and HLC

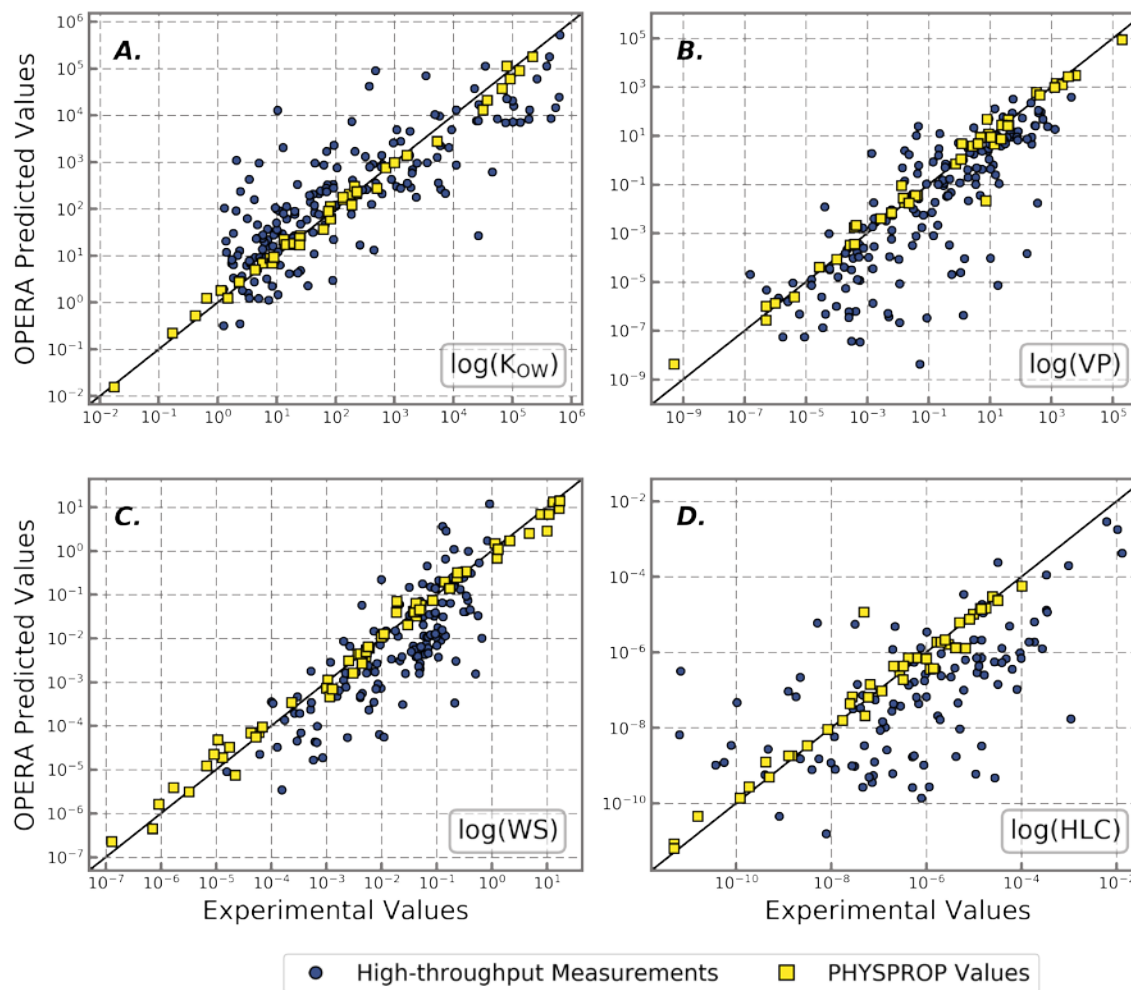
# Trends Across Modeling Tools

	N	RMSE					
		ACD/Labs	ChemProp*	EPI Suite	NICEATM	OPERA	PHYSPROP
<b>LogK<sub>ow</sub></b>	32	8.30E-01	7.91E-01	8.02E-01	7.75E-01	7.36E-01	7.71E-01
<b>VP</b>	33	-	1.44E+02	1.38E+02	-	1.36E+02	1.38E+02
<b>HLC</b>	23	-	-	5.99E-04	-	8.77E-05	8.96E-05
<b>WS</b>	36	1.20E+00	2.33E-01	3.07E-01	2.28E-01	5.07E-01	1.67E+00
<b>pKa<sup>†</sup></b>	76	6.15E+00	-	-	-	-	-

- Root mean squared errors (RMSE) are reported to emphasize average deviations between new values and previous or predicted values.
- The trends are as follows:
  - Log<sub>10</sub>K<sub>ow</sub>: OPERA < NICEATM < ChemProp < EPI Suite < ACD/Labs
  - VP: OPERA < EPI Suite < ChemProp
  - WS: ACD/Labs < NICEATM < ChemProp < EPI Suite < OPERA
  - HLC: OPERA < EPI Suite



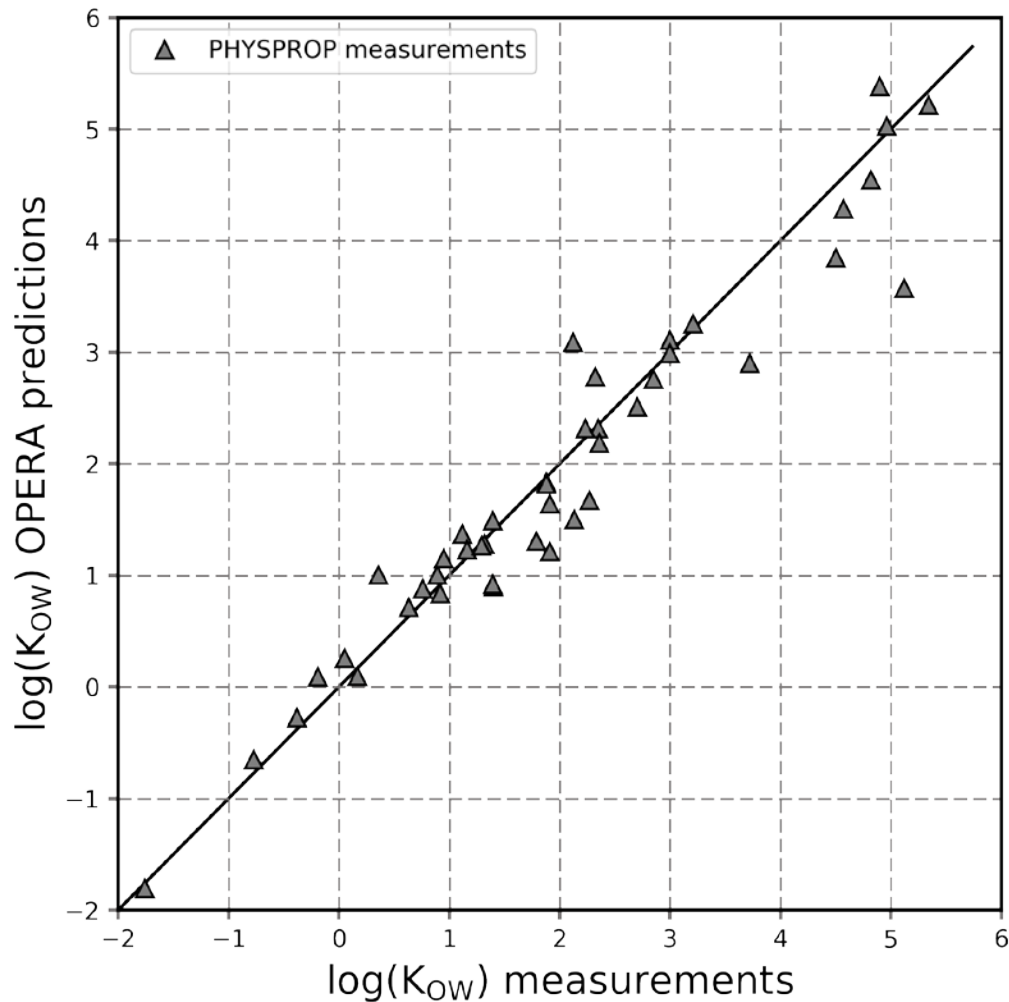
# OPERA Trends



- Comparison of OPERA predictions to both new and previous measurements

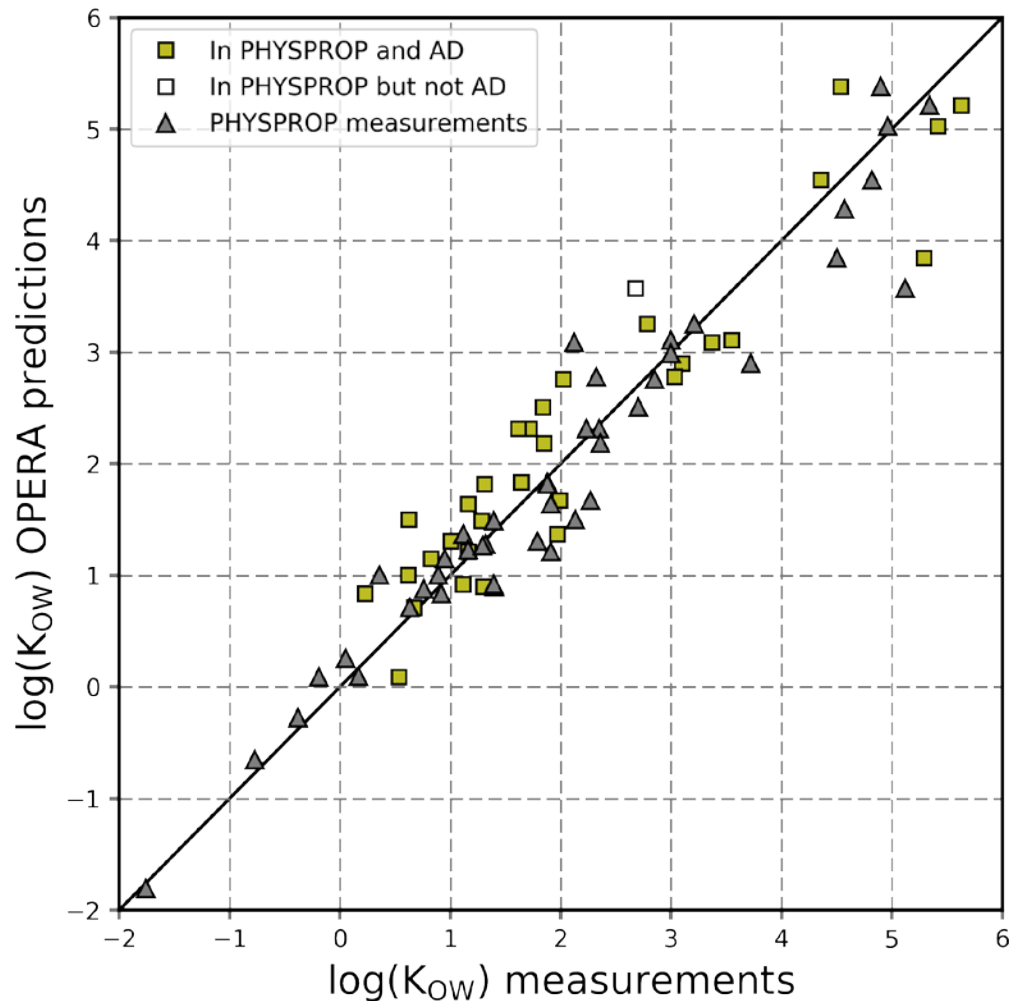
# OPERA $\log_{10}K_{ow}$ Trends and Applicability Domain

- Chemicals not in PHYSPROP were not in OPERA training set
- Chemicals with lower  $\log_{10}K_{ow}$  values were unamenable to new measurements



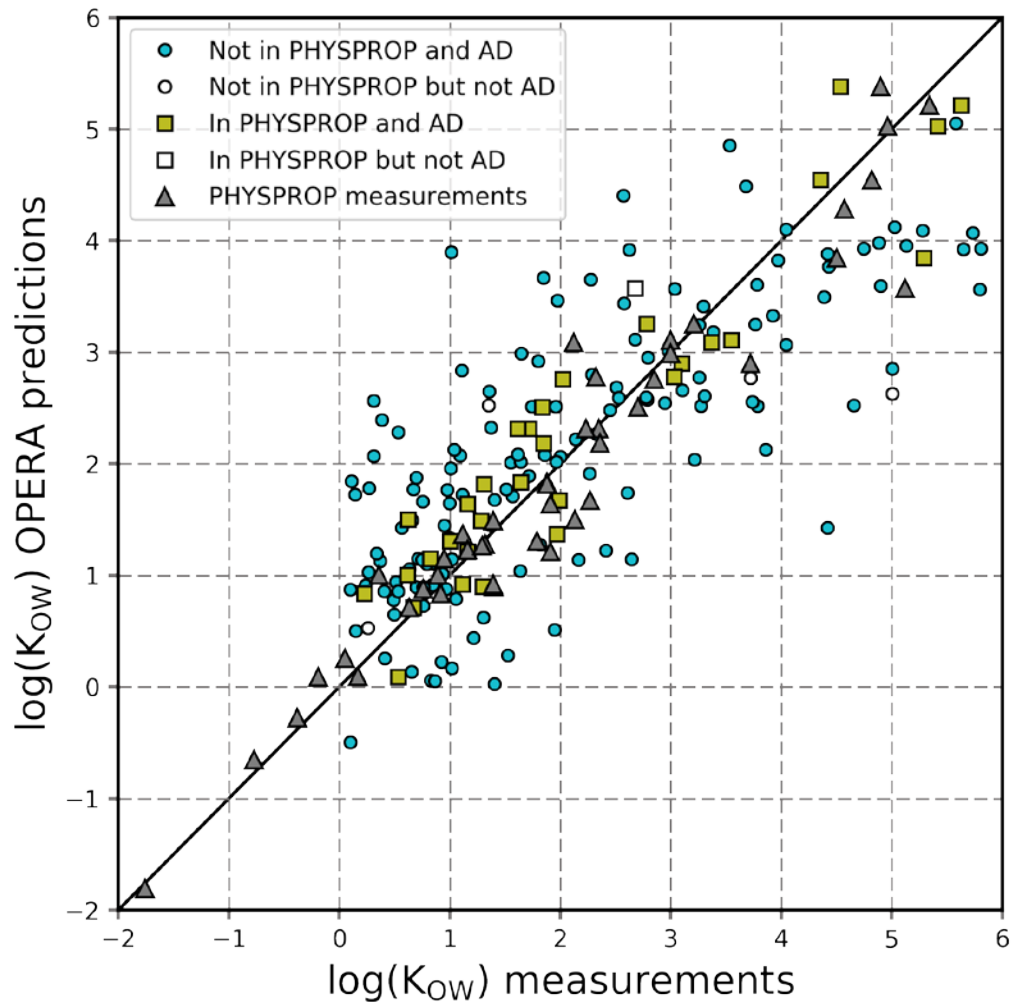
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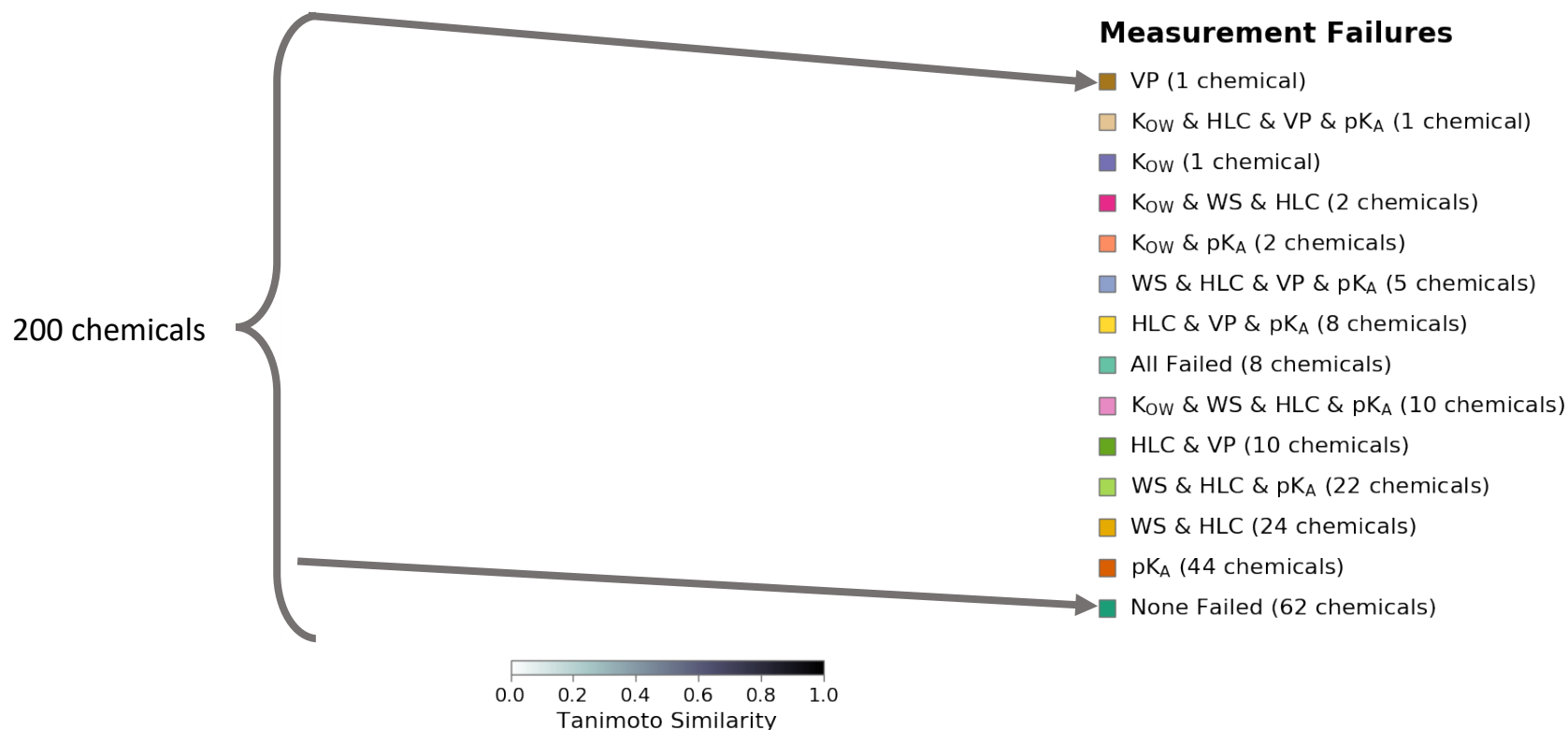


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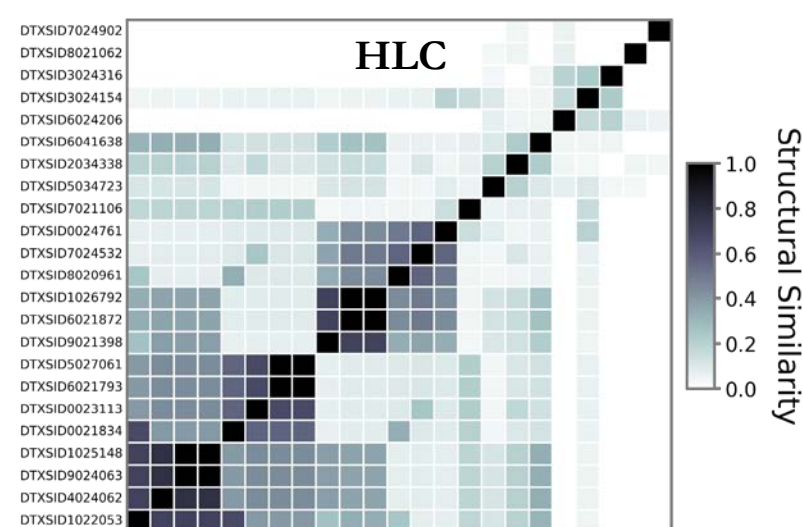
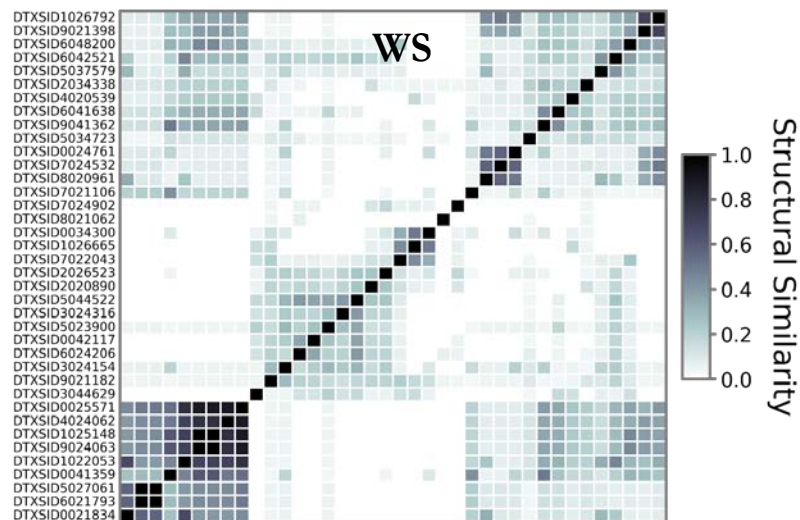
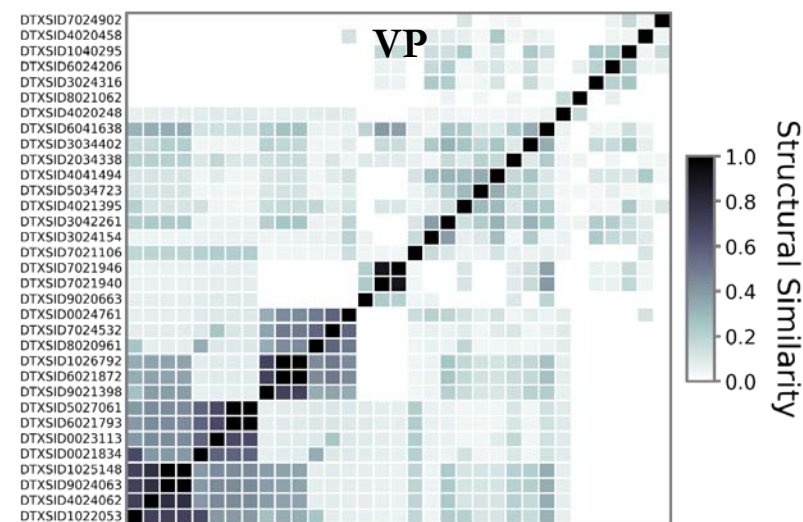
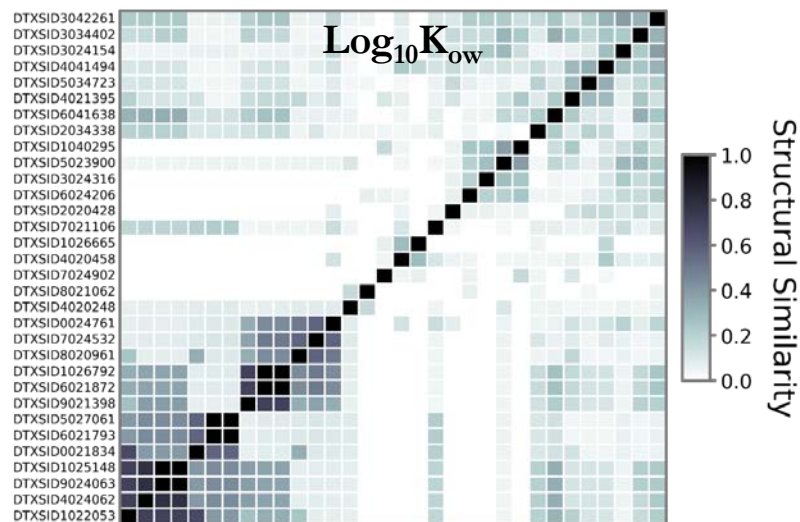


# Structure Similarity for Failures based by Property



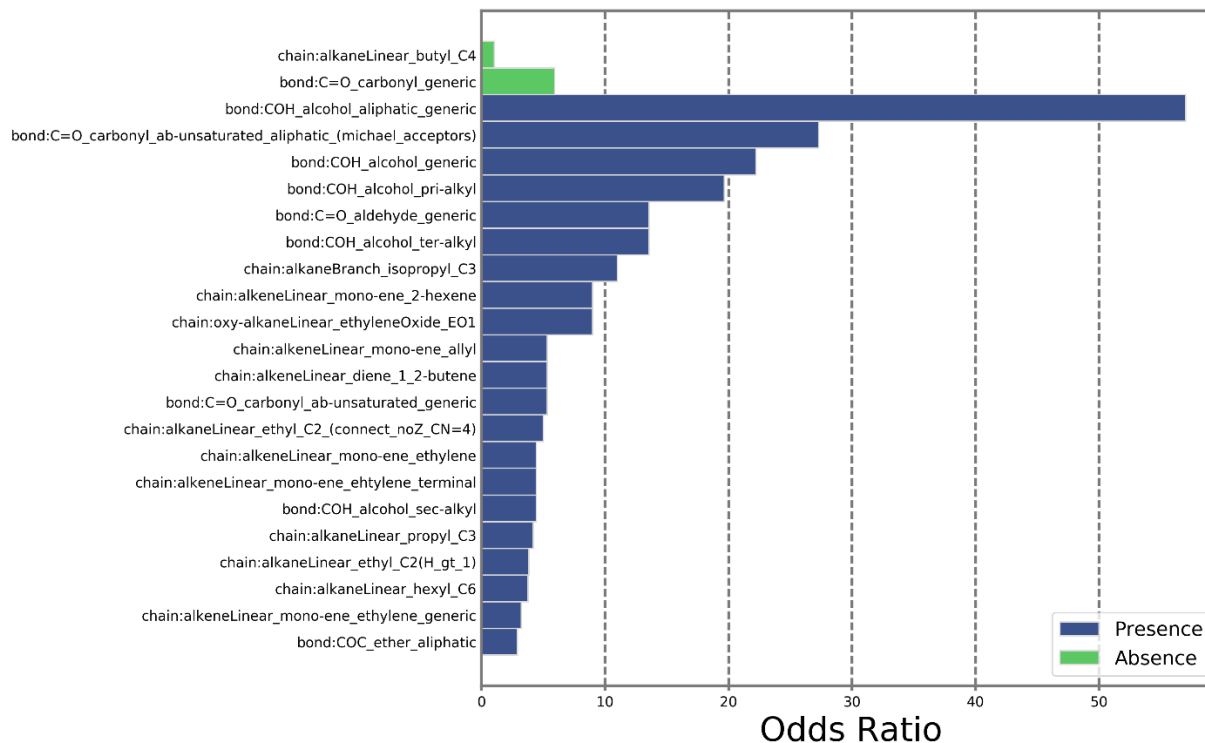
Chemotypes were used to seek patterns in measurement failures across properties

# Structural Similarity For New vs Previous Values



■ Structural similarity between chemicals from PHYSPROP and those with new measurements

# Odds Ratios for Successful Measurements



- Odds Ratio give the odds that a substructure contributes to failure of all measurements
  - Odds Ratio  $> 1 \rightarrow$  positive association
  - Odds Ratio  $= 1 \rightarrow$  no association
  - Odds Ratio  $< 1 \rightarrow$  negative association
- There were 21 substructures with positive associations
- There were 2 substructures with negative associations



# Conclusion

- Five physicochemical properties have been measured in a high-throughput fashion
- A selection process was implemented that allowed for diverse selection of 200 compounds for measurement
- Results from the new measurements were compared against both previously measured values and a collection of QSAR models that predict these values
- These new measurements can help improve chemical space covered by QSAR thus improving parameters for exposure and dose models



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*This work does not necessarily reflect U.S. EPA policy.*