



Deriving fish and *Daphnia* toxicity QSARs for anionic surfactants by using experimental and computational membrane-water partition coefficients

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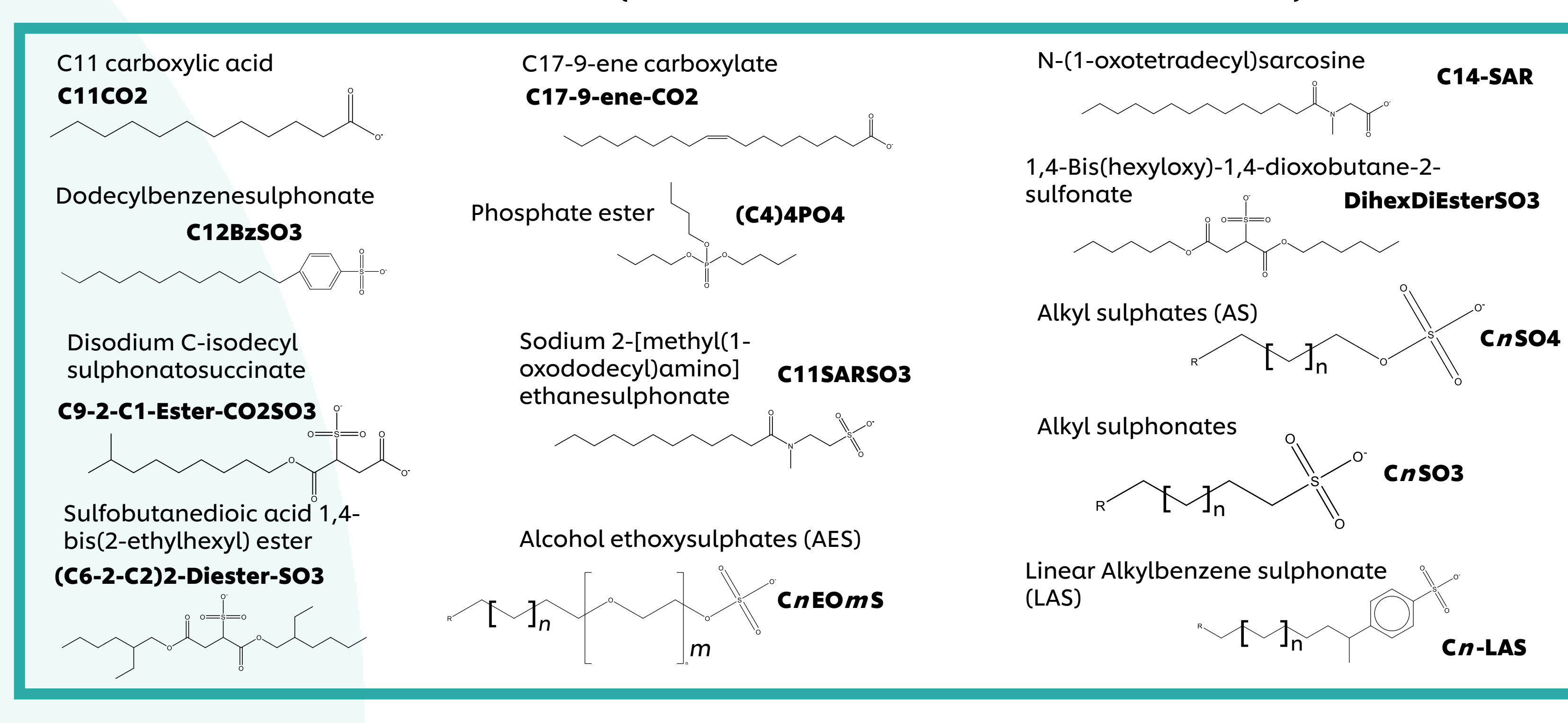
1. INTRODUCTION

- Quantitative Structure Activity Relationships (QSARs) are a viable alternative to *in-vivo* toxicity testing of chemicals.
- Many (eco)toxicity QSARs are hydrophobicity-based relationships using the octanol-water partition coefficient, $\log K_{OW}$ as a common descriptor for chemicals' toxicity.
- Determination of $\log K_{OW}$ for ionisable chemicals and surfactants is empirically difficult due to their tendency to accumulate at the octanol-water interface. Predictive methods are also often unreliable [1]
- Octanol cannot adequately describe the interactions of polar, charged, or amphiphilic compounds within ordered 3D structures of biological membranes.
- Membrane-water partition/distribution coefficient ($\log K_{MW}/D_{MW}$) provides a more biologically realistic approach for these compound types.
- As experimental determination of $\log K_{MW}/D_{MW}$ can also be complex and time consuming, we have calculated $\log K_{MW}/D_{MW}$ using the regression method developed by Droge *et al.* [2] and coarse-grained simulations [3] to develop fish and *Daphnia* toxicity QSARs for anionic surfactants.

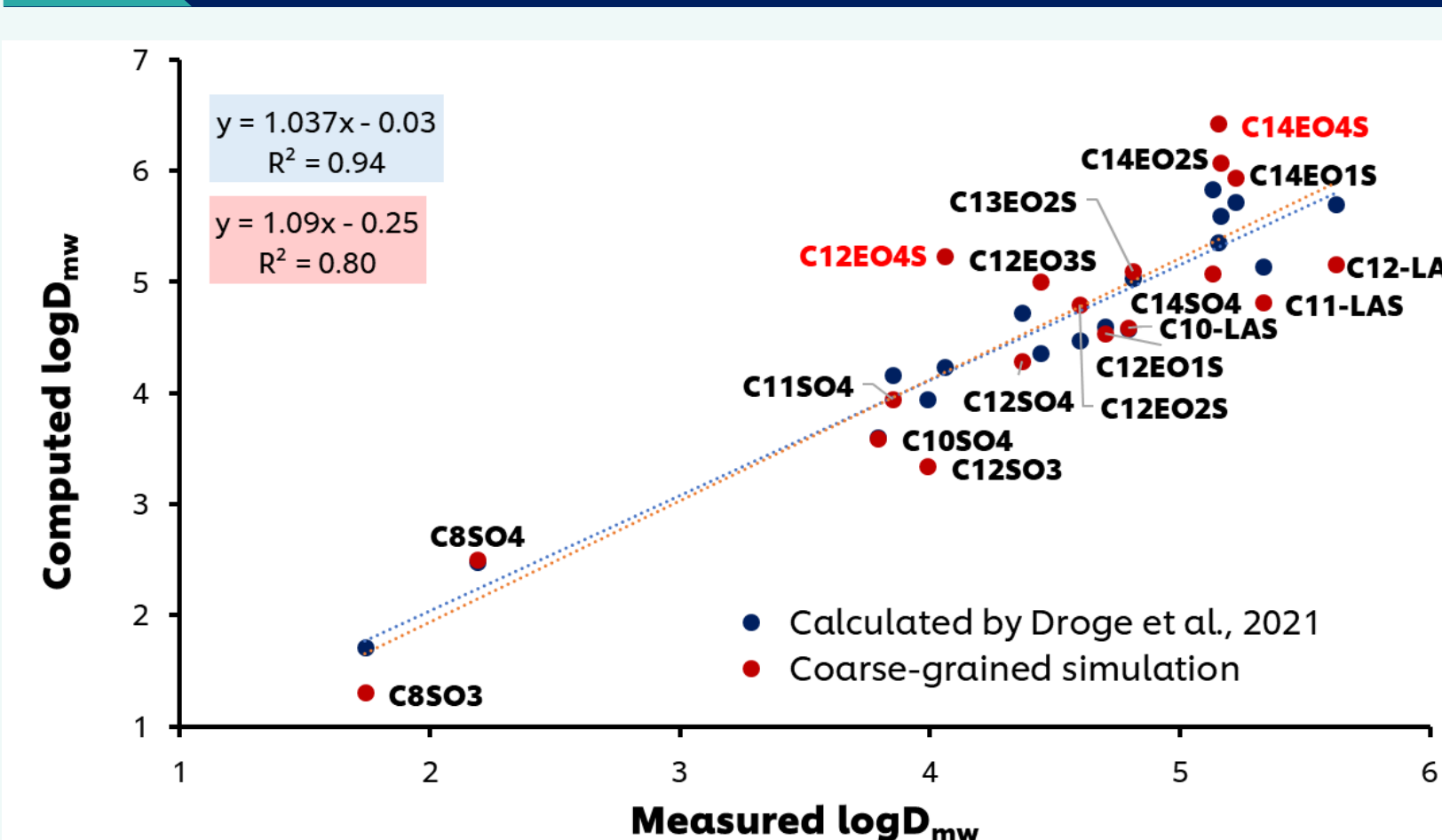
2. METHODOLOGY

- Previously, we have shown good correlation of simulated $\log D_{MW}$ against experimental values of chemicals, and presented several QSARs with homologue series of anionic surfactants successfully using $\log D_{MW}$ as an (eco)toxicity proxy [4,5].
- Here, database of fish and *Daphnia* literature toxicity data [6-8] containing mono constituent anionic surfactants (Table 1) was compiled and used to develop new QSARs.
- To address $\log D_{MW}$ experimental data gaps regarding the chemical space coverage (limited surfactant groups) we are also using two computational methods for $\log D_{MW}$ (Droge *et al.* [2] & coarse-grained simulation [3]), with the advantages and disadvantages of various methods for deriving $\log D_{MW}$ being discussed.
- Finally, we have used QSARs based on the same chemical space to make preliminary comparisons of species sensitivity for fish and *Daphnia*.

Table 1. List of used anionic surfactants (with chemical structures and abbreviations)



3. LogD_{MW} METHODS EVALUATION



Comparison of experimentally determined $\log D_{MW}$ with computational data indicates good correlation between the two.

- The Droge *et al.* method is a multiple regression based equation with limitations in surfactant group coverage (e.g. sarcosinates, phosphate esters), whereas simulation method is not restricted to certain surfactant groups.
- However, we have identified a limitation in simulation method for chemicals containing ethoxylate units (EO > 4).



Figure 1. Comparison of calculated $\log D_{MW}$ with measured $\log D_{MW}$ (SSLM method from Droge, 2019 [9] & Droge *et al.*, 2021 [3] & liposome internal data from SEAC, Unilever)

FISH (various test species)

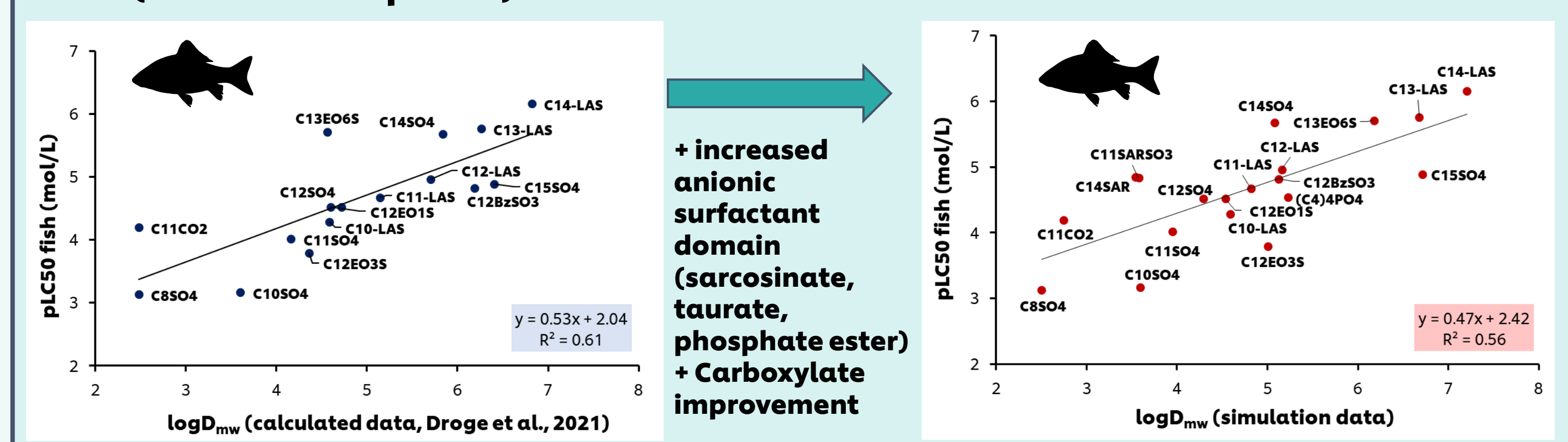


Figure 2. Comparison of fish QSARs based on calculated $\log D_{MW}$ by the Droge *et al.* regression and coarse-grained simulations

DAPHNIA (*D. magna*, *D. pulex*, *Ceriodaphnia dubia*)

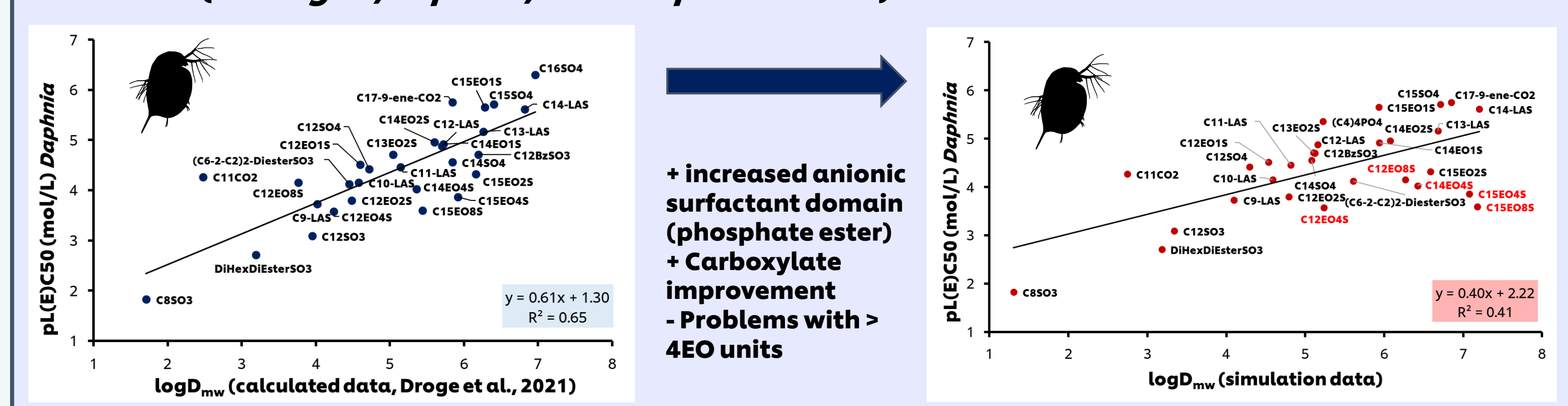


Figure 3. Comparison of *Daphnia* QSARs based on calculated $\log D_{MW}$ by the Droge *et al.* regression and coarse-grained simulations

4. FISH & DAPHNIA logD_{MW} QSARs

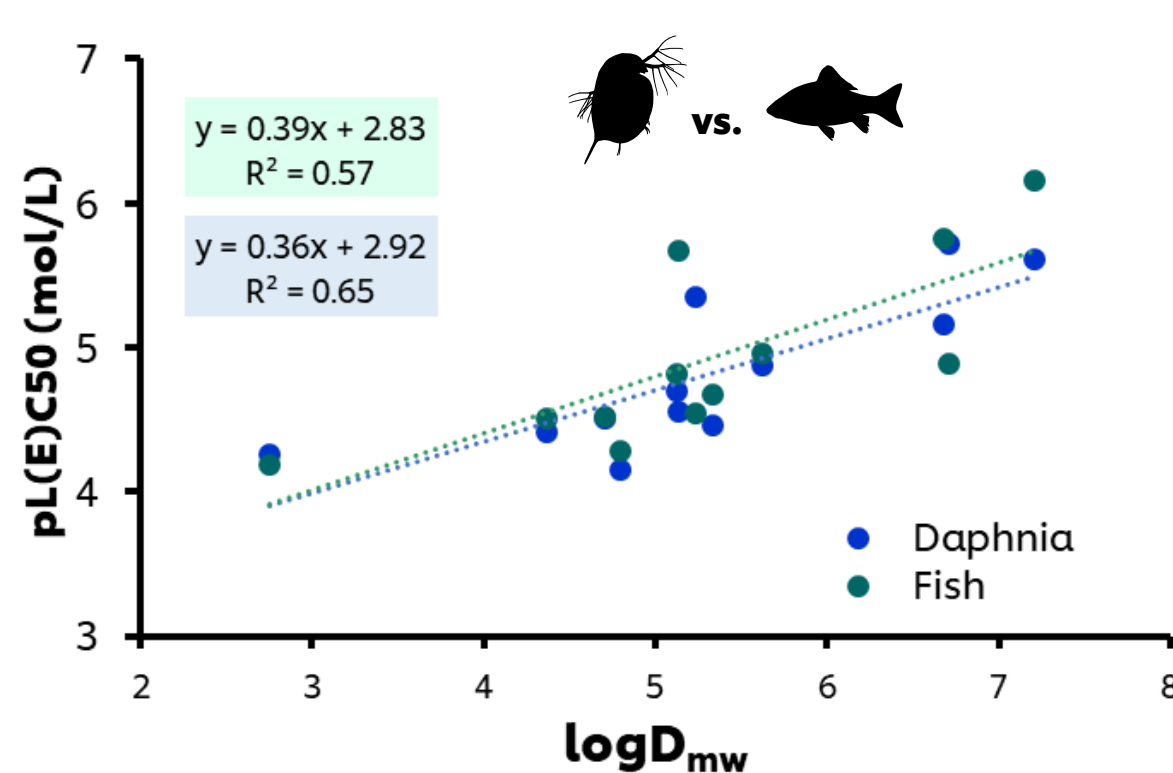
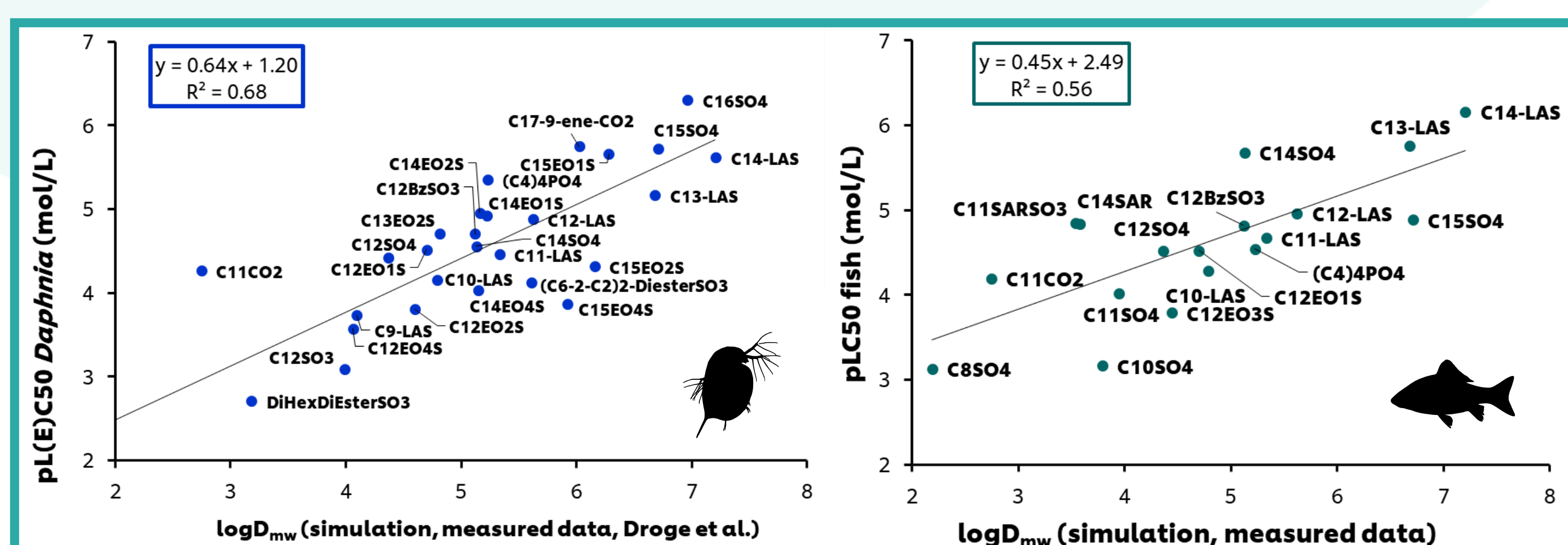


Figure 5. Fish and *Daphnia* QSARs comparison

- QSARs were compared for chemicals where both *Daphnia* and fish toxicity data are available (including LAS, AS, AES, phosphate esters).
- Statistically significant differences were not observed between slopes and intercepts, indicating there is **potentially no difference between sensitivity** of these trophic levels to anionic surfactants.

Figure 4. Anionic surfactants fish and *Daphnia* QSARs

- This work represents the first step towards generating $\log D_{MW}$ based QSARs for anionic surfactants with potential to be used in **environmental risk assessment** and/or **regulatory submissions** avoiding the need for unnecessary fish testing



5. CONCLUSIONS AND FUTURE WORK

- Using literature (eco)toxicity data for a variety of mono constituent anionic surfactants we have shown it possible to generate **general anionic surfactant - based QSAR** using $\log D_{MW}$ for the prediction of aquatic toxicity to fish and *Daphnia*.
- Species sensitivity** between fish and *Daphnia* based on the same chemical dataset suggests that there is no difference between the two. This could provide the opportunity to waive the need for acute fish toxicity testing for anionic surfactants. However, further work would be required to expand the data set to provide a more robust analysis.

Development of a reliable computational $\log D_{MW}$ method which covers a wide chemical space is required. Further work has been identified to **refine the approach for chemicals with EO > 4**.

To improve the QSAR regression, consideration could be given to **include additional (eco)toxicity data from multi-constituent surfactants**, however this requires previous knowledge of the chain length distribution in order to predict a representative $\log D_{MW}$.

6. REFERENCES

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