

Experiences of Predicting Acute Oral Toxicity in Industry

Richard Marchese Robinson, Surya Hari Lal, Sarah Whalley
Syngenta, Jealott's Hill International Research Centre, Bracknell, RG42 6EY, UK

richard.marchese_robinson@syngenta.com

Overview

- Knowledge of **rodent acute oral toxicity (AOT)** is critical for assessing the **safety** and **registerability** of plant protection product active ingredients (AIs)
- Ethical imperative to reduce animal tests and extensive testing of AI leads in early stage research is not feasible
- Predictive models based on the (Quantitative) Structure-Activity Relationship [(Q)SAR] paradigm can support early stage research and may support future regulatory submissions

Early Stage Research

- Models can be used to
 - perform *in silico* screens, guiding projects away from problematic regions of chemical space
 - prioritise compounds for synthesis and testing
- Global** models support predictions across a range of projects
- Local** models are project specific – tuned to an AI lead and its derivatives

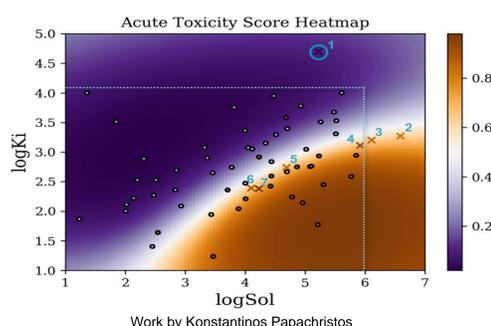


Regulatory Submissions?

- Legislation in different regions supports the reduction (or elimination) of animal testing for regulatory approval of (certain) chemicals^{1,2}
- EU and US scientific agencies have recently expressed an interest in alternatives to *in vivo* AOT testing for plant protection products – but significant barriers remain to acceptance³
- The US EPA's commitment² to eliminate mammal studies by 2035 may provide further impetus for the acceptance of (Q)SAR predictions of AOT to support regulatory submissions

In-House Local Model Example

- Estimated posterior probability of acute toxicity at a single dose is plotted against model inputs characterising affinity for the site of toxicological action (predicted/measured K_i) and bioavailability (predicted/measured solubility)
- Compounds with low AOT probability prioritised for progression – avoiding late stage attrition
- However, building this kind of model requires a single, *known* mode of toxicological action

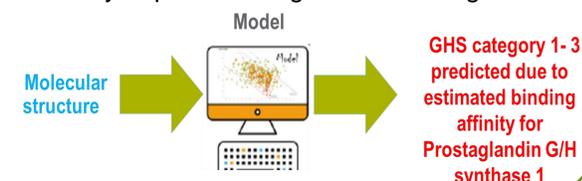


Evaluation of External Global Models

- Various Open Source and commercial software programs now provide (Q)SAR models to predict the AOT LD₅₀ or predict a category corresponding to a range of LD₅₀ values, e.g. GHS categories
 - Leadscope,⁴ OPERA, UL Cheminformatics Toolkit etc.
- A variety of software programs and models were identified for an evaluation of their ability to categorise compounds according to critical LD₅₀ thresholds on a curated Syngenta in-house AOT dataset
 - Do our early stage research compounds lie inside the applicability domains of these models?
 - Which model is best?
 - Is the best model good enough to support early stage research projects?

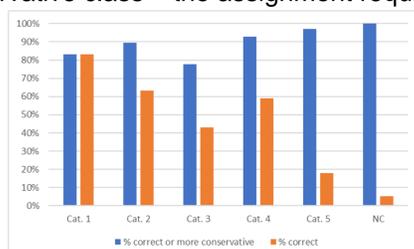
In-House Global Model?

- The option of building an in-house global model is under consideration
- This could improve the chance of our AI leads lying inside the applicability domain
- The model could be iteratively updated with new data
 - Mechanistic interpretability is desired to
 - guide AI molecular design
 - support possible future regulatory acceptance
 - One option – adapt a recently published framework,⁵ combining molecular descriptors and predicted protein-binding affinities for a variety of possible targets of toxicological action

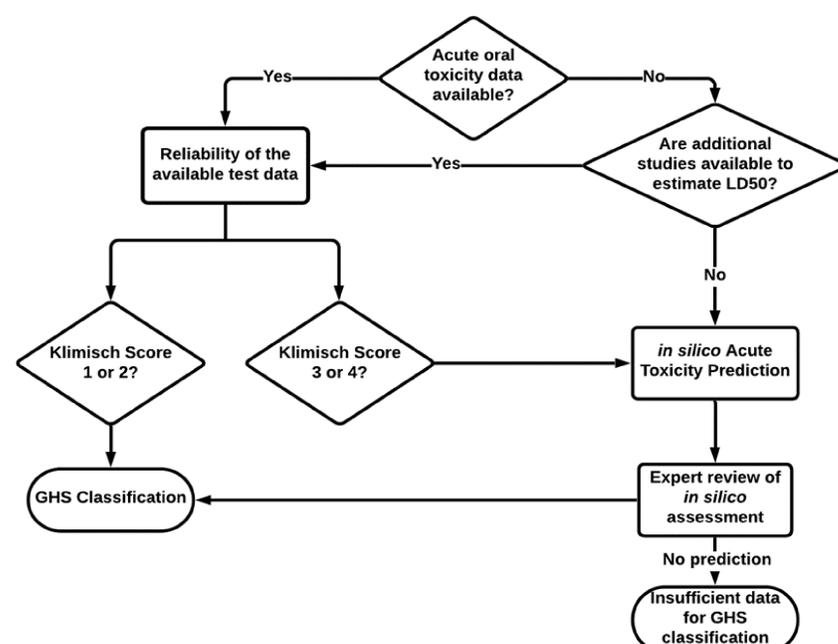


Preliminary Leadscope Results⁴

- 657 chemicals (Syngenta, PPDB)⁶ assigned GHS categories 1 – 5, or non-classified (NC), using rat AOT test data
 - Where possible, restricted to acceptable quality data (Klimisch score = 1 or 2)
 - Highly unbalanced – only 6 compounds in GHS category 1 (LD₅₀ ≤ 5 mg/kg)
- Multi-class predictions made using consensus of Partial Logistic Regression and structural alert models
 - The most conservative (lowest GHS category) prediction is returned
 - The results are encouraging
 - Only 21 inconclusive predictions had to be discarded
 - The balanced accuracy (45%) is much higher than would be expected due to chance (17%)
- On average, 90% of chemicals in a given category are assigned to the correct or a more conservative class – the assignment required in a regulatory setting



Possible Workflow to Support Regulatory Submissions⁴



References

- Negro et al. Regul. Toxicol. Pharmacol. 2018, 99, 33-49.
- <https://www.epa.gov/research/administrator-memo-prioritizing-efforts-reduce-animal-testing-september-10-2019>
- Prior et al. Regul. Toxicol. Pharmacol. 2019, 102, 30-33.
- Bercu et al. A cross-industry collaboration to assess if acute oral toxicity (Q)SAR models are fit-for-purpose for GHS classification and labelling. (Manuscript in Preparation)
- Allen et al. J. Cheminform. 2019, 11, 36.
- Lewis et al. Hum. Ecol. Risk Assess, 2016, 22, 1050-1064.